

Precision Stellar Photometry and Astrometry using Discrete Point Spread Functions: Theory and Practice

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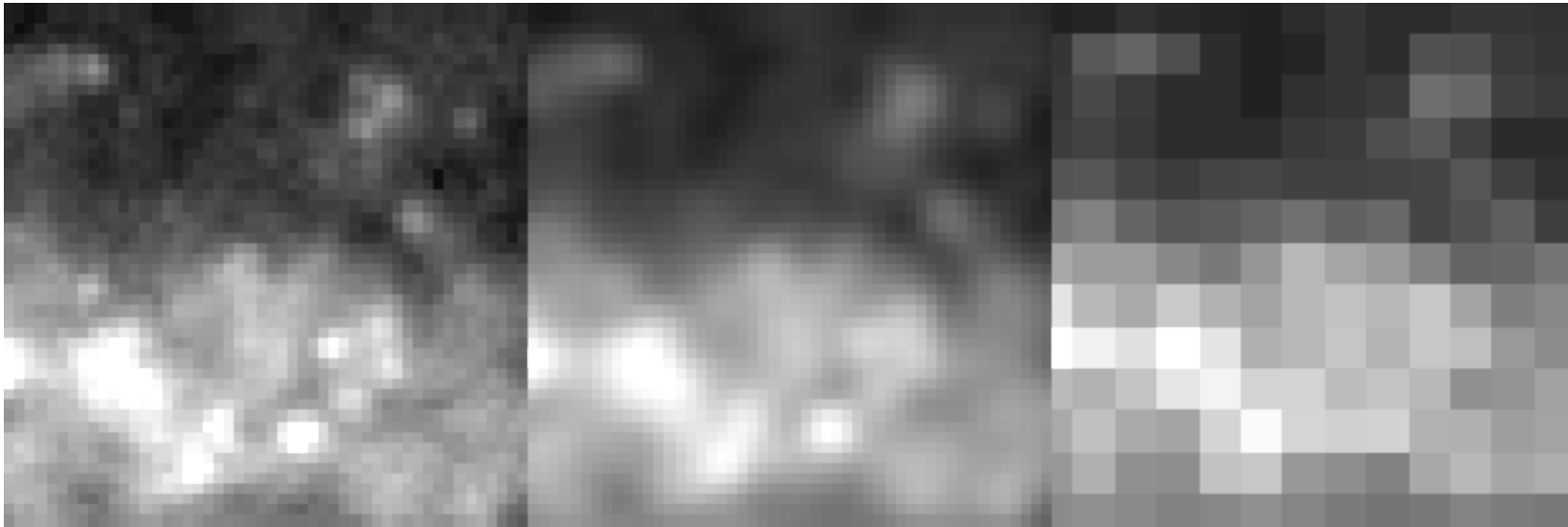


Outline

- Introduction
- Point Response Functions
- Performance Model based on Information Theory
- Photometry and Astrometry with Discrete PSFs
- Spitzer Space Telescope's Infrared Array Camera

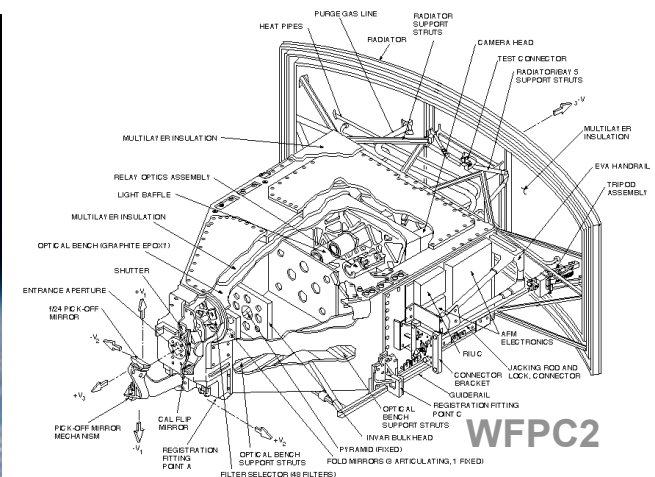
Introduction

Image Formation Process



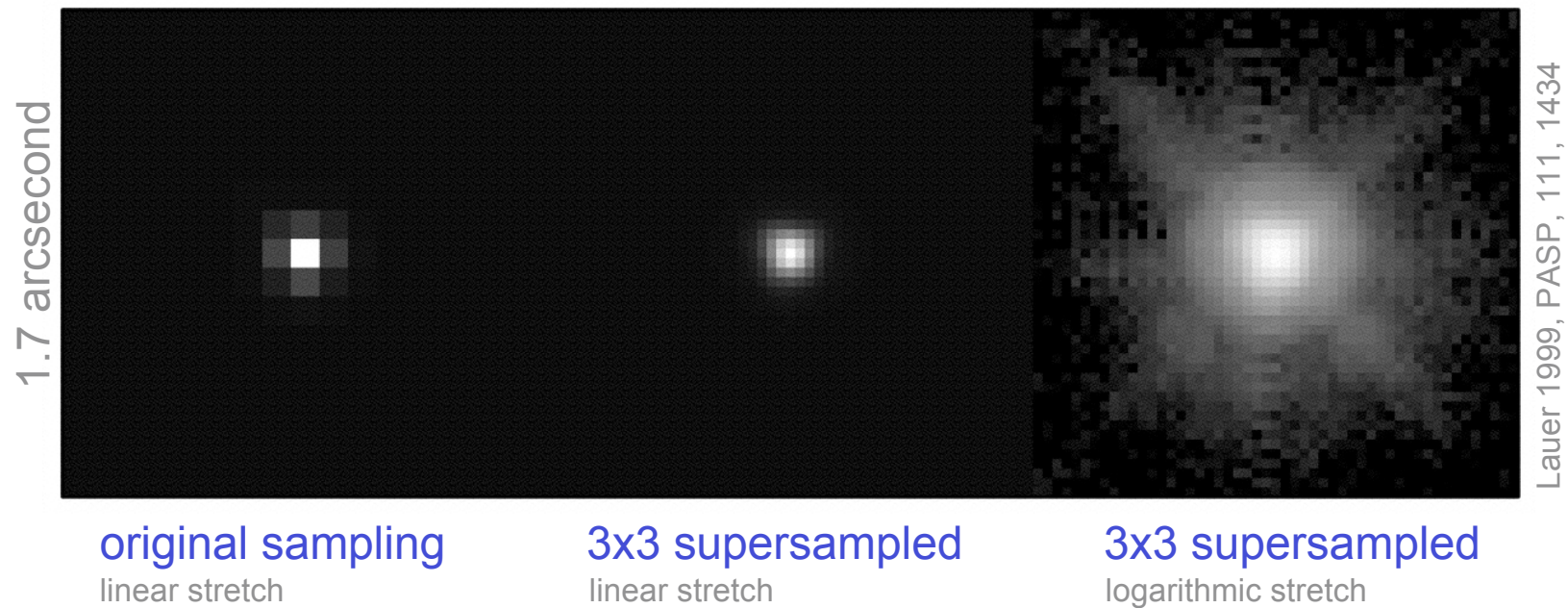
Fruchter & Hook 2000 (ADASS IX)

SKY \Rightarrow TELESCOPE \Rightarrow DETECTOR



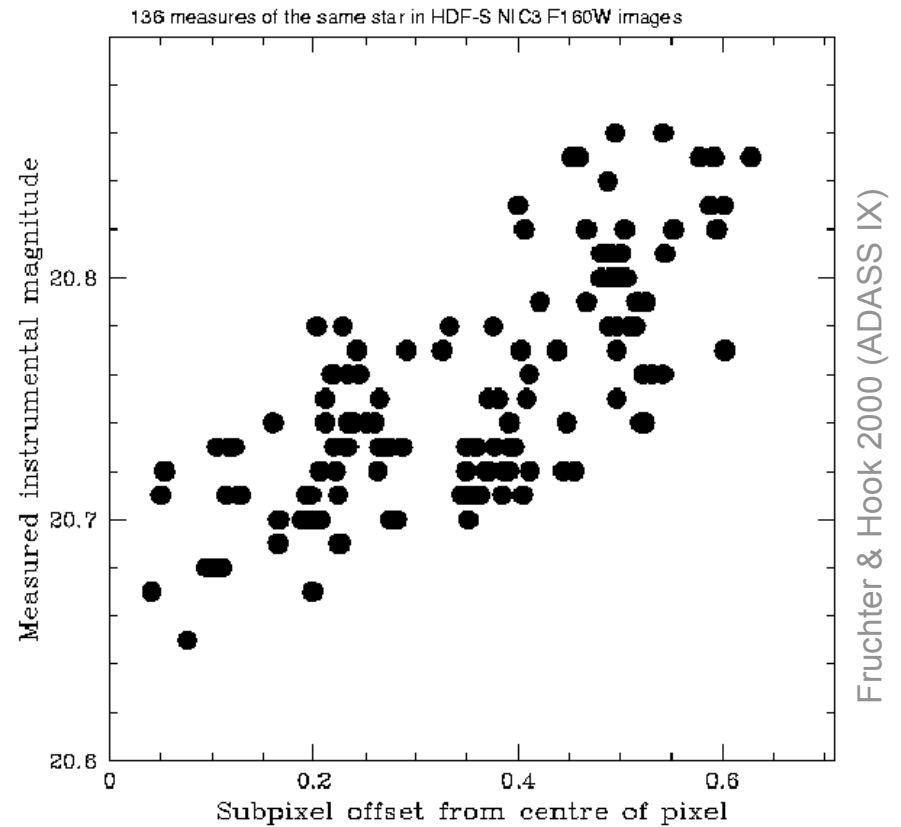
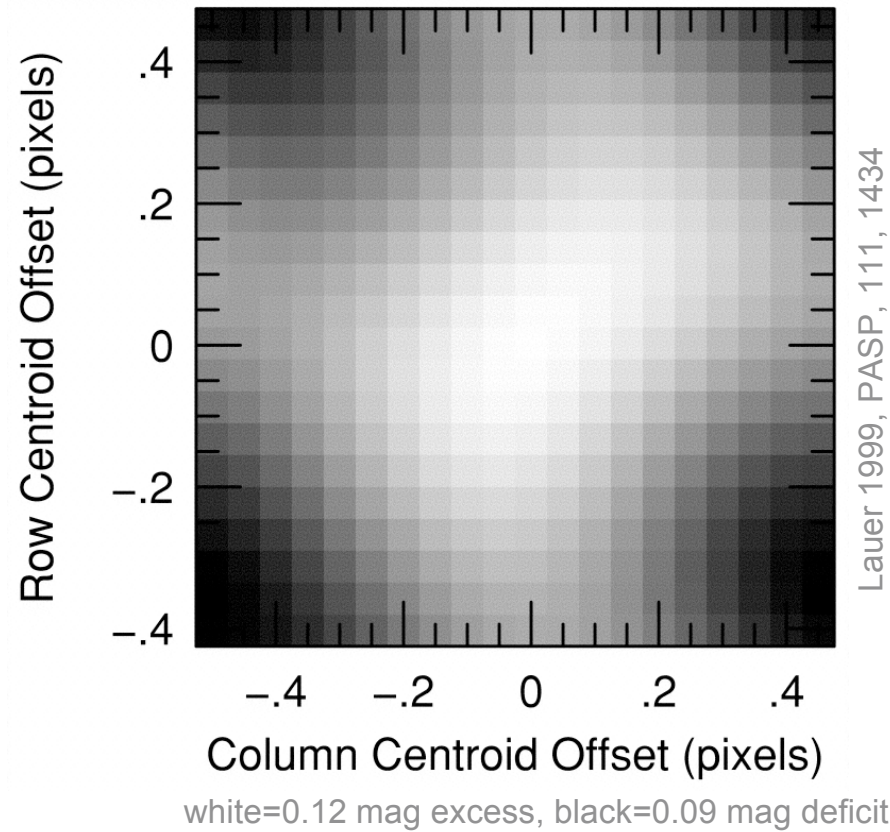
Undersampled Point Spread Functions

HST WFPC2: WFC F555W 0.10 arcsecond pixels

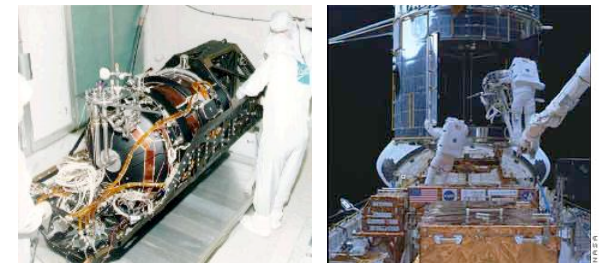


Lossy Detectors

Photometric Error: NIC3 F160W



NICMOS



Point Response Functions

Point Response Function ▼

▼ Point Spread Function

$$\Psi \equiv \phi * \Lambda$$

▲ Detector Response Function

$$\Psi_i(x_i, y_i) \equiv \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \phi(x, y) dx dy \quad (\text{for an ideal DRF})$$

$$V \equiv \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \Psi dx dy \lesssim 1$$

Volume ▲

$$\text{sharpness} \equiv \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left(\frac{\Psi}{V} \right)^2 dx dy$$

▲ normalized PRF

$$\beta \equiv \left[\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \Psi^2 dx dy \right]^{-1} = \frac{1}{V^2 \text{sharpness}}$$

Effective Background Area ▲

$$\mathcal{L} \equiv \sqrt{\frac{\beta V^2}{4\pi}} = \frac{1}{\sqrt{4\pi \text{sharpness}}}$$

Critical-Sampling Scale Length ▲

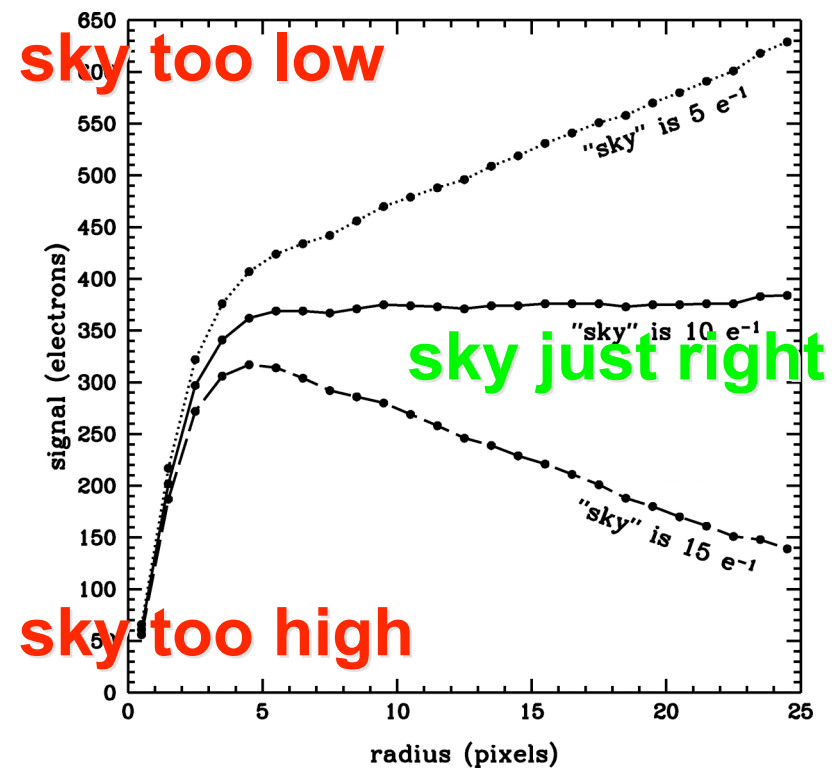
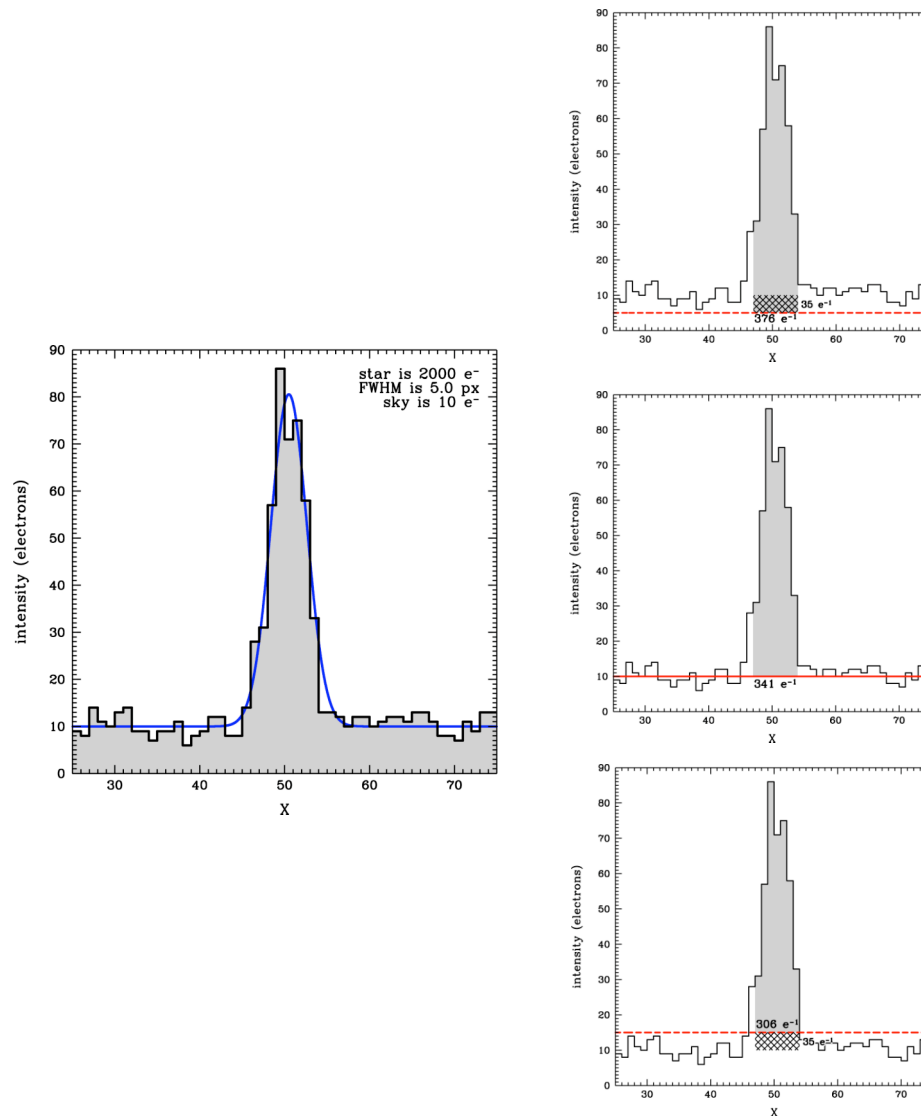
Performance Model based on Information Theory

I. Photometry

$$\begin{aligned}
\sigma_{\mathcal{E}: \text{bright}}^2 &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial \mathcal{E}} \right)^2 \right]^{-1} \\
&\approx \left[\sum_{i=1}^N \frac{1}{\mathcal{E} V \tilde{\Psi}_i} \left(\frac{\partial}{\partial \mathcal{E}} \left[\mathcal{E} V \tilde{\Psi}_i + \mathcal{B} \right] \right)^2 \right]^{-1} \\
&= \left[\frac{1}{\mathcal{E} V} \sum_{i=1}^N \frac{1}{\tilde{\Psi}_i} \left(V \tilde{\Psi}_i \right)^2 \right]^{-1} \\
&= \left[\frac{V}{\mathcal{E}} \sum_{i=1}^N \tilde{\Psi}_i \right]^{-1} \\
&\approx \frac{\mathcal{E}}{V} \left[\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \tilde{\Psi} \, dx \, dy \right]^{-1} \\
&\equiv \frac{\mathcal{E}}{V}
\end{aligned}$$

$$\begin{aligned}
\sigma_{\text{rms}} &\equiv \sqrt{\frac{1}{N} \sum_{i=1}^N \sigma_i^2} \approx \sqrt{\mathcal{B} + \sigma_{\text{RON}}^2} \\
\sigma_{\mathcal{E}:\text{faint}}^2 &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial \mathcal{E}} \right)^2 \right]^{-1} \\
&\approx \left[\sum_{i=1}^N \frac{1}{\sigma_{\text{rms}}^2} \left(\frac{\partial}{\partial \mathcal{E}} [\mathcal{E} V \tilde{\Psi}_i + \mathcal{B}] \right)^2 \right]^{-1} \\
&\equiv \sigma_{\text{rms}}^2 \left[\sum_{i=1}^N (\Psi_i)^2 \right]^{-1} \\
&\approx \sigma_{\text{rms}}^2 \left[\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \Psi^2 dx dy \right]^{-1} \\
&\equiv \beta \sigma_{\text{rms}}^2 \\
&\approx \beta [\mathcal{B} + \sigma_{\text{RON}}^2]
\end{aligned}$$

Systematic Error: Poor Sky Measurement



$$\sigma_{\mathcal{E}: \text{faint}} \approx \sigma_{\text{rms}} \sqrt{\beta} + \underline{\sigma_{\mathcal{B}}} \beta$$

$$\begin{aligned} \sigma_{\mathcal{B}} &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial \mathcal{B}} \right)^2 \right]^{-1/2} \\ &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_{\text{rms}}^2} \left(\frac{\partial}{\partial \mathcal{B}} [\mathcal{E} V \Psi_i + \mathcal{B}] \right)^2 \right]^{-1/2} \\ &= \sigma_{\text{rms}} \left[\sum_{i=1}^N (1)^2 \right]^{-1/2} \\ &= \frac{\sigma_{\text{rms}}}{\sqrt{N}} \\ &\approx \sqrt{\frac{\mathcal{B} + \sigma_{\text{RON}}^2}{N}} \end{aligned}$$

$$\begin{aligned}
\sigma_{\mathcal{E}: \text{faint}} &\approx \sigma_{\text{rms}} \sqrt{\beta} + \sigma_{\mathcal{B}} \beta \\
&= \sqrt{\beta} \left(1 + \sqrt{\beta/N} \right) \sigma_{\text{rms}} \\
&\approx \sqrt{\beta} \left(1 + \sqrt{\beta/N} \right) \sqrt{\mathcal{B} + \sigma_{\text{RON}}^2}
\end{aligned}$$

$$\begin{aligned}
S/N &\lesssim \frac{\mathcal{E}}{\sigma_{\mathcal{E}}} \\
&\approx \frac{\mathcal{E}}{\sqrt{\sigma_{\mathcal{E}:\text{bright}}^2 + \sigma_{\mathcal{E}:\text{faint}}^2}} \\
&\approx \frac{\mathcal{E}}{\sqrt{\frac{\mathcal{E}}{V} + \beta \left(1 + \sqrt{\beta/N}\right)^2 \sigma_{\text{rms}}^2}} \\
&\approx \frac{\mathcal{E}}{\sqrt{\frac{\mathcal{E}}{V} + \beta \left(1 + \sqrt{\beta/N}\right)^2 [\mathcal{B} + \sigma_{\text{RON}}^2]}}
\end{aligned}$$

$$\Delta\text{mag} = \frac{5 / \ln(100)}{\text{SNR}} = \frac{2.5 \log(e)}{\text{SNR}} \approx \frac{1.0857}{S/N}$$

II. Astrometry

$$\tilde{g}(x, y; \mathcal{X}, \mathcal{Y}, \mathcal{S}) \equiv \frac{1}{2\pi\mathcal{S}^2} e^{-\left[\frac{(x - \mathcal{X})^2 + (y - \mathcal{Y})^2}{2\mathcal{S}^2}\right]}$$

$$\tilde{g}_i \equiv \tilde{g}(x_i, y_i; \mathcal{X}, \mathcal{Y}, \mathcal{S})$$

$$\tilde{G}_i \equiv \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \tilde{g}(x, y; \mathcal{X}, \mathcal{Y}, \mathcal{S}) dx dy$$

$$\tilde{G}_i \approx \tilde{g}_i \quad (\text{for } \mathcal{S} \gg 1)$$

$$m_i \equiv \mathcal{E}V\tilde{G}_i + \mathcal{B}$$

$$\begin{aligned}
\sigma_{\mathcal{X}:\text{bright}}^2 &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial \mathcal{X}} \right)^2 \right]^{-1} \\
&\approx \left[\sum_{i=1}^N \frac{1}{\mathcal{E}V\tilde{\mathbf{G}}_i} \left(\frac{\partial}{\partial \mathcal{X}} \left[\mathcal{E}V\tilde{\mathbf{G}}_i + \mathcal{B} \right] \right)^2 \right]^{-1} \\
&\approx \left[\sum_{i=1}^N \frac{1}{\mathcal{E}V\tilde{\mathbf{g}}_i} \left(\mathcal{E}V \frac{\partial}{\partial \mathcal{X}} \tilde{\mathbf{g}}_i \right)^2 \right]^{-1} \\
&= \left[\sum_{i=1}^N \frac{1}{\mathcal{E}V\tilde{\mathbf{g}}_i} \left(\mathcal{E}V \tilde{\mathbf{g}}_i \frac{x_i - \mathcal{X}}{\mathcal{S}^2} \right)^2 \right]^{-1} \\
&\approx \frac{\mathcal{S}^4}{\mathcal{E}V} \left[\iint_{-\infty}^{+\infty} \tilde{\mathbf{g}}(x, y; \mathcal{X}, \mathcal{Y}, \mathcal{S}) (x - \mathcal{X})^2 dx dy \right]^{-1} \\
&= \frac{\mathcal{S}^2}{\mathcal{E}V} \\
&\approx \frac{\mathcal{L}^2}{\mathcal{E}V}, \text{ where } \mathcal{L} \equiv \sqrt{\frac{\beta V^2}{4\pi}} \equiv \sqrt{\frac{\tilde{\beta}}{4\pi}} \equiv \frac{1}{\sqrt{4\pi} \text{ sharpness}} \\
&\quad \quad \quad \blacktriangle = 1 \text{ if critically sampled}
\end{aligned}$$

$$\begin{aligned}
\sigma_{\mathcal{X}:\text{faint}}^2 &\approx \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial \mathcal{X}} \right)^2 \right]^{-1} \\
&\approx \left[\sum_{i=1}^N \frac{1}{\sigma_{\text{rms}}^2} \left(\frac{\partial}{\partial \mathcal{X}} \left[\varepsilon V \tilde{\mathbf{G}}_i + \mathcal{B} \right] \right)^2 \right]^{-1} \\
&\approx \sigma_{\text{rms}}^2 \left[\sum_{i=1}^N \left(\varepsilon V \frac{\partial}{\partial \mathcal{X}} \tilde{\mathbf{g}}_i \right)^2 \right]^{-1} \\
&= \sigma_{\text{rms}}^2 \left[\sum_{i=1}^N \left(\varepsilon V \tilde{\mathbf{g}}_i \frac{x_i - \mathcal{X}}{S^2} \right)^2 \right]^{-1} \\
&\approx \sigma_{\text{rms}}^2 \frac{S^4}{\varepsilon^2 V^2} \left[\iint_{-\infty}^{+\infty} [\tilde{\mathbf{g}}(x, y; \mathcal{X}, \mathcal{Y}, S)(x - \mathcal{X})]^2 dx dy \right]^{-1} \\
&= 8 \pi \sigma_{\text{rms}}^2 \frac{S^4}{\varepsilon^2 V^2} \\
&\approx 8 \pi \sigma_{\text{rms}}^2 \left(\frac{\mathcal{L}^2}{\varepsilon V} \right)^2 \approx 8 \pi \sigma_{\text{rms}}^2 \left(\sigma_{\mathcal{X}:\text{bright}}^2 \right)^2
\end{aligned}$$

$$\begin{aligned}
\sigma_{\mathcal{X}} &\approx \sqrt{\sigma_{\mathcal{X}:\text{bright}}^2 + \sigma_{\mathcal{X}:\text{faint}}^2} \\
&\approx \sqrt{\frac{\mathcal{L}^2}{\mathcal{E} V} \left[1 + 8 \pi \sigma_{\text{rms}}^2 \frac{\mathcal{L}^2}{\mathcal{E} V} \right]} \\
&\approx \sqrt{\frac{\mathcal{L}^2}{\mathcal{E} V} \left[1 + 8 \pi (\mathcal{B} + \sigma_{\text{RON}}^2) \frac{\mathcal{L}^2}{\mathcal{E} V} \right]}
\end{aligned}$$

$$\sigma_{\mathcal{Y}} = \sigma_{\mathcal{X}}$$

III. Relationship Between Astrometric and Photometric Errors

Bright stars:

$$\frac{\sigma_{\mathcal{X}}}{\mathcal{L}} = \frac{\sigma_{\mathcal{E}}}{\mathcal{E}}$$

Faint stars:

$$\frac{\sigma_{\mathcal{X}}}{\mathcal{L}} \approx \left(\frac{\sigma_{\mathcal{E}}}{\mathcal{E}} \right) \frac{\sqrt{2}}{1 + \sqrt{\beta/N}}$$

IV. Performance Model

background level ▼

▼ volume of Point Response Function

$$m_i \equiv \mathcal{B} + \mathcal{E}V\tilde{\Psi}_i$$

pixel value ▲
stellar intensity [electrons] ▲

▲ normalized PRF

$$\tilde{\Psi}_i(x_i, y_i) \equiv \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \phi(x, y) dx dy \quad (\text{ideal DRF: } V \equiv 1)$$

▲ Point Spread Function

$$\text{S/N} \equiv \frac{\mathcal{E}}{\sigma_{\mathcal{E}}} \approx \frac{\mathcal{E}}{\sqrt{\frac{\mathcal{E}}{V} + \beta \left(1 + \sqrt{\beta/N}\right)^2 [\mathcal{B} + \sigma_{\text{RON}}^2]}}$$

signal-to-noise ratio ▲

▲ readout noise

▲ aperture size [pixels]

Effective Background Area

$$\Delta \text{mag} \approx \frac{1.0857}{\text{S/N}}$$

▼ Critical-Sampling Scale Length

$$\sigma_{\mathcal{X}} \approx \sqrt{\frac{\mathcal{L}^2}{\mathcal{E}V} \left[1 + 8\pi (\mathcal{B} + \sigma_{\text{RON}}^2) \frac{\mathcal{L}^2}{\mathcal{E}V} \right]}$$

$$\sigma_{\mathcal{Y}} = \sigma_{\mathcal{X}}$$

Mon. Not. R. Astron. Soc. **361**, 861–878 (2005)

doi:10.1111/j.1365-2966.2005.09208.x

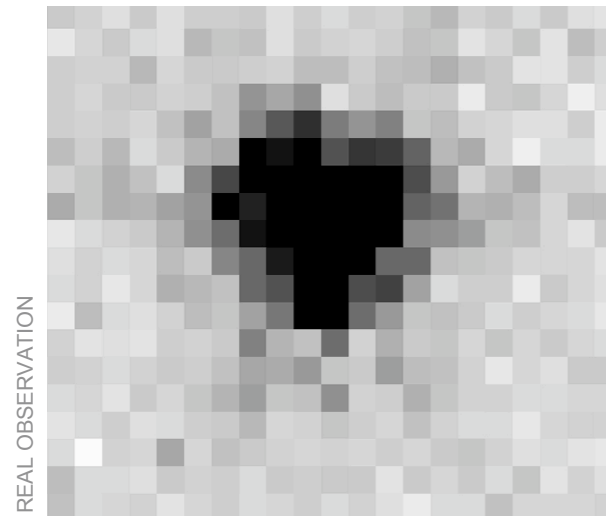
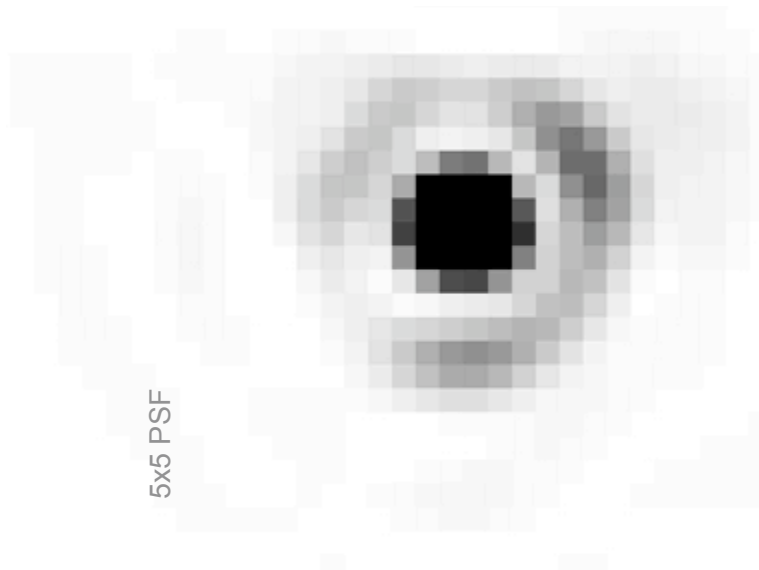
Stellar photometry and astrometry with discrete point spread functions

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Photometry and Astrometry with Discrete PSFs

What is a discrete PSF?



Infrared Array Camera



Spitzer Space Telescope



$$\chi^2(\mathbf{p}) \equiv \sum_{i=1}^N \frac{1}{\sigma_i^2} (z_i - m_i)^2$$

parameter vector \blacktriangle \blacktriangle data \blacktriangle model \blacktriangle measurement error

$$\chi^2(\mathbf{p} + \boldsymbol{\delta}) = \sum_{n=0}^{\infty} \frac{1}{n!} (\boldsymbol{\delta} \cdot \nabla)^n \chi^2(\mathbf{p})$$

$$\approx \chi^2(\mathbf{p}) + \boldsymbol{\delta} \cdot \nabla \chi^2(\mathbf{p}) + \frac{1}{2} \boldsymbol{\delta} \cdot \mathbf{H} \cdot \boldsymbol{\delta}$$

\blacktriangle Hessian matrix

$$[\mathbf{H}]_{jk} \equiv \frac{\partial^2}{\partial \mathbf{a}_j \partial \mathbf{a}_k} \chi^2(\mathbf{p})$$

$$\mathbf{H} \cdot \boldsymbol{\delta} = -\nabla \chi^2(\mathbf{p})$$

$$\mathbf{p}' = \mathbf{p} + \boldsymbol{\delta}$$

$$\sigma_j \approx \sqrt{[\mathbf{H}^{-1}]_{jj}} = \left[\sum_{i=1}^N \frac{1}{\sigma_i^2} \left(\frac{\partial m_i}{\partial p_j} \right)^2 \right]^{-1/2}$$

measurement error \blacktriangle

How does one move a PSF?

Analytical PSFs: Just compute the PSF at the desired location in the observational model.

Numerical PSFs: Take the reference numerical PSF and shift it to the desired location using a perfect 2-d interpolation function. OK... but how is that done in practice?

Solution

Use the following 21-pixel-wide damped sinc function:

$$f^{\text{shifted}}(\delta x) \equiv \sum_{i=-10}^{10} f(x_i) \frac{\sin\left(\pi(x_i - \delta x)\right)}{\pi(x_i - \delta x)} \exp\left(-\frac{(x_i - \delta x)^2}{(3.25)^2}\right)$$

Note: The 2-d sinc function is separable in x and y .

Hessian matrix

Standard definition for model $\mathbf{m} = \boldsymbol{\varepsilon} \mathbf{V} \tilde{\Psi}(\boldsymbol{\chi}, \boldsymbol{\gamma}) + \boldsymbol{\mathcal{B}}$ and N data values:

$$\mathbf{H} \equiv \begin{bmatrix} \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\varepsilon} \partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\varepsilon} \partial \boldsymbol{\chi}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\varepsilon} \partial \boldsymbol{\gamma}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\varepsilon} \partial \boldsymbol{\mathcal{B}}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\chi} \partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\chi} \partial \boldsymbol{\chi}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\chi} \partial \boldsymbol{\gamma}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\chi} \partial \boldsymbol{\mathcal{B}}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\gamma} \partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\gamma} \partial \boldsymbol{\chi}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\gamma} \partial \boldsymbol{\gamma}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\gamma} \partial \boldsymbol{\mathcal{B}}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\mathcal{B}} \partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\mathcal{B}} \partial \boldsymbol{\chi}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\mathcal{B}} \partial \boldsymbol{\gamma}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial^2 m_i}{\partial \boldsymbol{\mathcal{B}} \partial \boldsymbol{\mathcal{B}}} \end{bmatrix}$$

Hessian matrix

Robust approximation for model $\mathbf{m} = \boldsymbol{\varepsilon} \mathbf{V} \tilde{\Psi}(\mathbf{x}, \mathbf{y}) + \mathbf{B}$ and N data values:

$$H \approx \begin{bmatrix} \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} \frac{\partial m_i}{\partial \mathbf{x}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} \frac{\partial m_i}{\partial \mathbf{y}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} \frac{\partial m_i}{\partial \mathbf{B}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{x}} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{x}} \frac{\partial m_i}{\partial \mathbf{x}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{x}} \frac{\partial m_i}{\partial \mathbf{y}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{x}} \frac{\partial m_i}{\partial \mathbf{B}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{y}} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{y}} \frac{\partial m_i}{\partial \mathbf{x}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{y}} \frac{\partial m_i}{\partial \mathbf{y}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{y}} \frac{\partial m_i}{\partial \mathbf{B}} \\ \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{B}} \frac{\partial m_i}{\partial \boldsymbol{\varepsilon}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{B}} \frac{\partial m_i}{\partial \mathbf{x}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{B}} \frac{\partial m_i}{\partial \mathbf{y}} & \sum_{i=1}^N \frac{1}{\sigma_i^2} \frac{\partial m_i}{\partial \mathbf{B}} \frac{\partial m_i}{\partial \mathbf{B}} \end{bmatrix}$$

Jacobian matrix

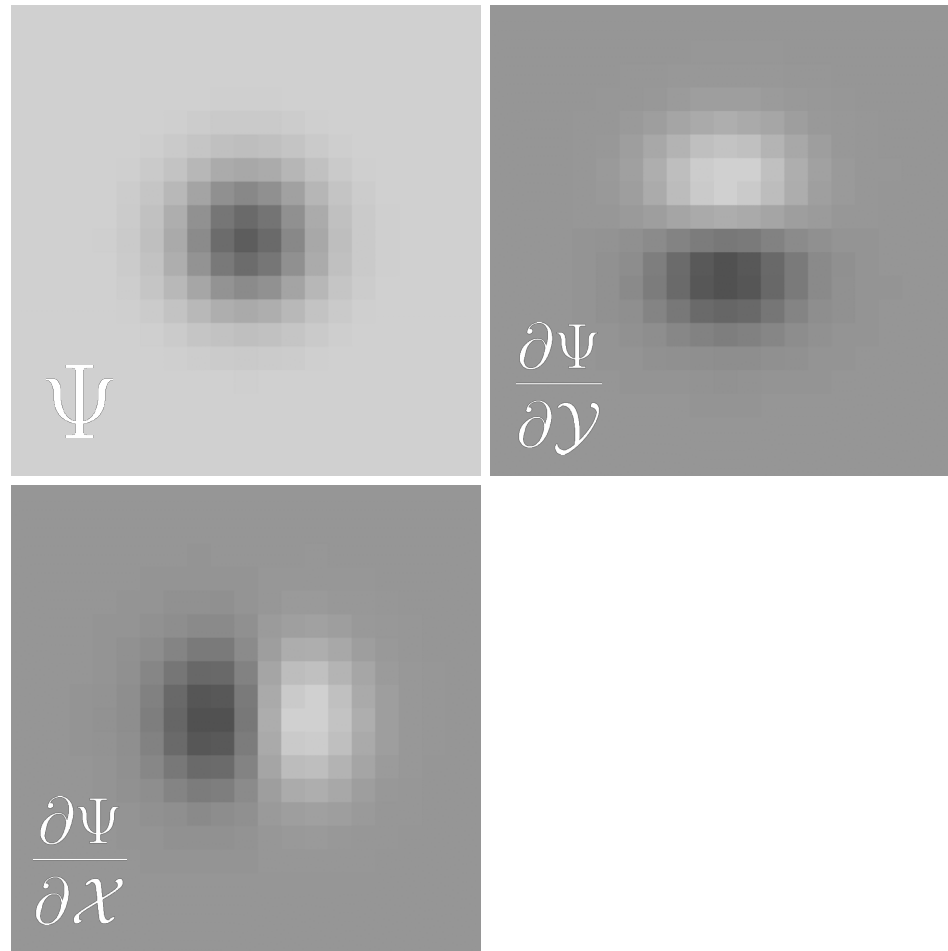
For model $m_i \equiv \mathcal{E} \vee \tilde{\Psi}_i(\mathcal{X}, \mathcal{Y}) + \mathcal{B}$ with N data values,

$$J \equiv \begin{bmatrix} \frac{\partial m_1}{\partial \mathcal{E}} & \frac{\partial m_1}{\partial \mathcal{X}} & \frac{\partial m_1}{\partial \mathcal{Y}} & \frac{\partial m_1}{\partial \mathcal{B}} \\ \frac{\partial m_2}{\partial \mathcal{E}} & \frac{\partial m_2}{\partial \mathcal{X}} & \frac{\partial m_2}{\partial \mathcal{Y}} & \frac{\partial m_2}{\partial \mathcal{B}} \\ \frac{\partial m_3}{\partial \mathcal{E}} & \frac{\partial m_3}{\partial \mathcal{X}} & \frac{\partial m_3}{\partial \mathcal{Y}} & \frac{\partial m_3}{\partial \mathcal{B}} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{\partial m_N}{\partial \mathcal{E}} & \frac{\partial m_N}{\partial \mathcal{X}} & \frac{\partial m_N}{\partial \mathcal{Y}} & \frac{\partial m_N}{\partial \mathcal{B}} \end{bmatrix}$$

▲ ▲ ▲ ▲

Each column forms an image

Derivatives of the PRF

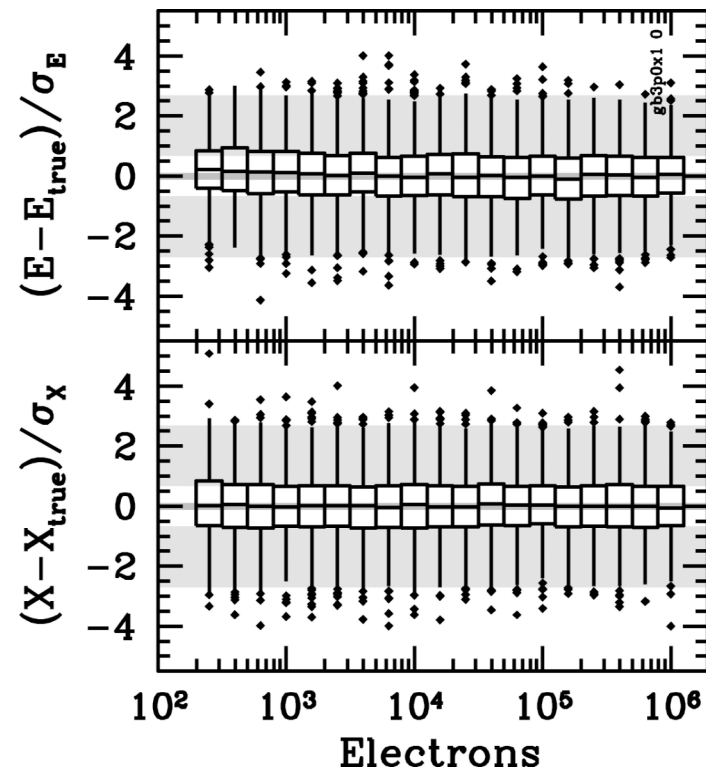
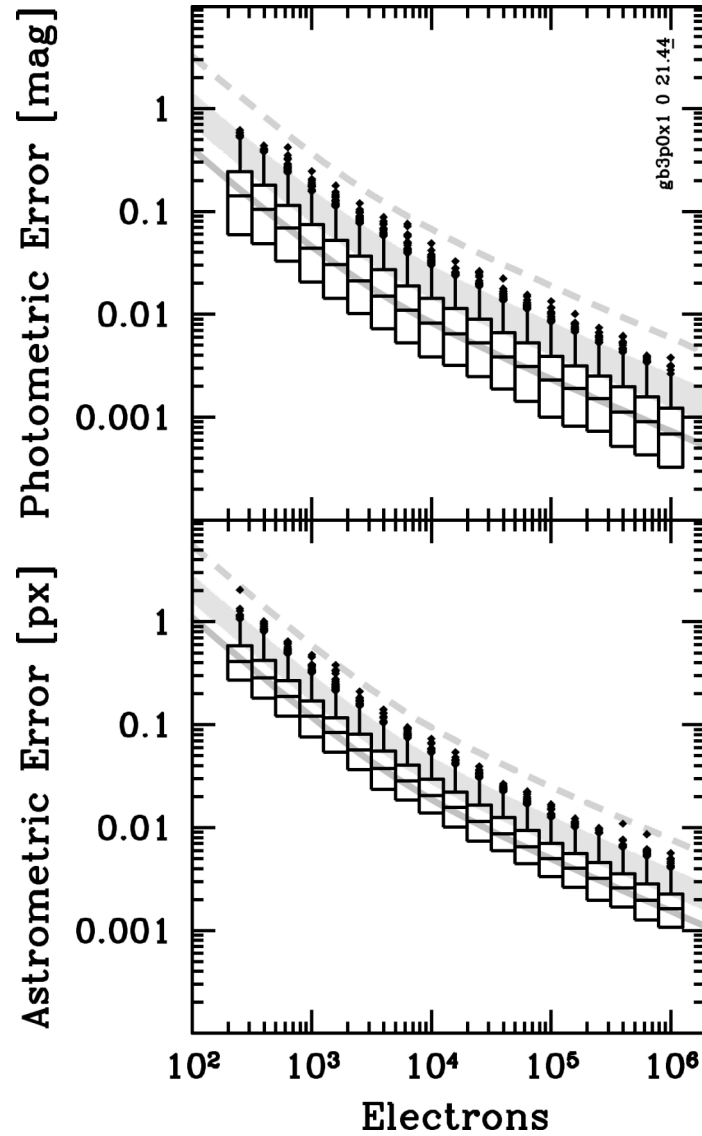


Analytical Derivatives

$$\begin{aligned}\Psi_i(x_i, y_i; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S}) &\equiv \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \phi(x, y; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S}) dx dy \quad (\text{ideal DRF}) \\ &\approx \frac{1}{\eta^2} \sum_{j=1}^{\eta} \sum_{k=1}^{\eta} \phi(x_i - \Delta + j\delta, y_i - \Delta + k\delta; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S}) \\ &\text{where } \Delta \equiv \frac{\eta + 1}{2\eta} \quad \text{and} \quad \delta \equiv \frac{1}{\eta}\end{aligned}$$

$$\begin{aligned}\frac{\partial}{\partial \mathcal{X}} \Psi_i &\equiv \frac{\partial}{\partial \mathcal{X}} \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \phi(x, y; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S}) dx dy \quad (\text{ideal DRF}) \\ &\approx \frac{1}{\eta^2} \sum_{j=1}^{\eta} \sum_{k=1}^{\eta} \frac{\partial}{\partial \mathcal{X}} \phi(x_i - \Delta + j\delta, y_i - \Delta + k\delta; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S})\end{aligned}$$

$$\begin{aligned}\frac{\partial}{\partial \mathcal{Y}} \Psi_i &\equiv \frac{\partial}{\partial \mathcal{Y}} \int_{x_i-0.5}^{x_i+0.5} \int_{y_i-0.5}^{y_i+0.5} \phi(x, y; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S}) dx dy \quad (\text{ideal DRF}) \\ &\approx \frac{1}{\eta^2} \sum_{j=1}^{\eta} \sum_{k=1}^{\eta} \frac{\partial}{\partial \mathcal{Y}} \phi(x_i - \Delta + j\delta, y_i - \Delta + k\delta; \mathcal{E}, \mathcal{X}, \mathcal{Y}, \mathcal{S})\end{aligned}$$

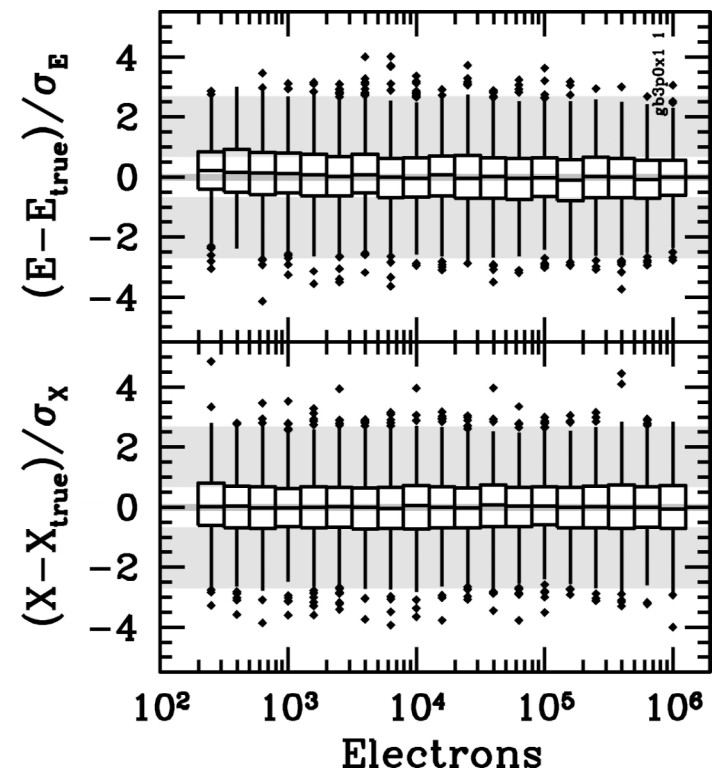
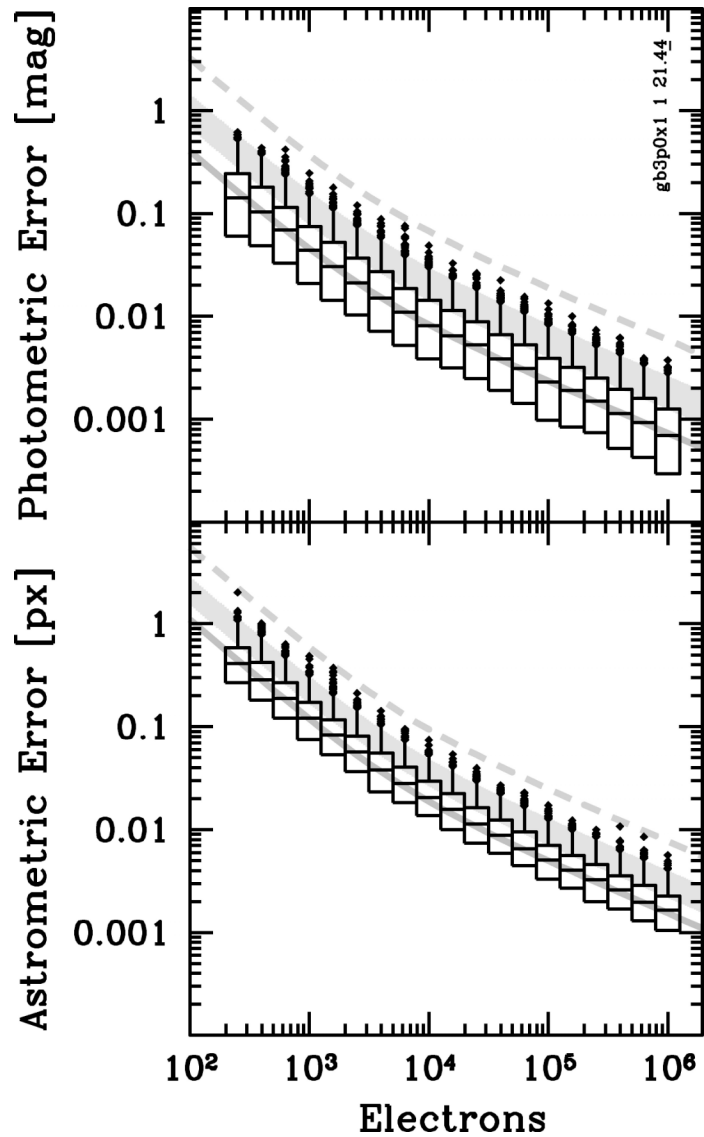


Data: Analytical Gaussian FWHM=3.0 px
 Model: Analytical Gaussian FWHM=3.0 px

Numerical Derivatives

The **mathematics** of determining the position partial derivatives of the observational model with respect to the x and y direction vectors **is exactly the same with analytical or numerical PSFs**. **The implementation methodology, however, is significantly different.** The position partial derivatives of numerical PSFs can be determined using **numerical differentiation techniques** on the numerical PSF. Numerical experiments have shown that the following five-point differentiation formula works well with numerical PSFs:

$$f'(x_i) \approx \frac{1}{12} \left[f(x_{i-2}) - 8f(x_{i-1}) + 8f(x_{i+1}) - f(x_{i+2}) \right]$$

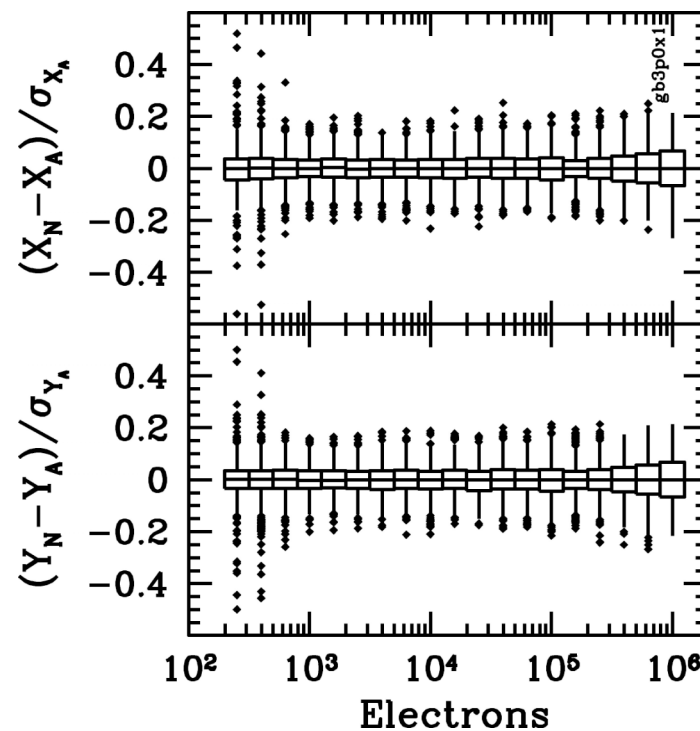


Data: Analytical Gaussian FWHM=3.0 px
 Model: Numerical Gaussian FWHM=3.0 px

Why does this work so well?

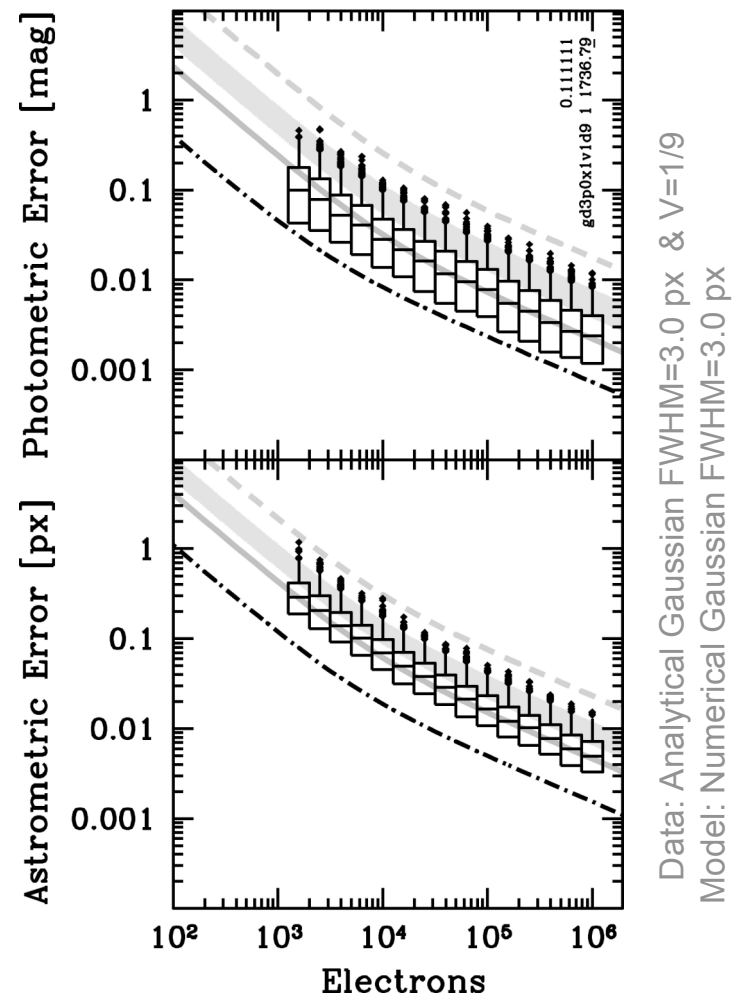
Photon Noise

▲and That's A Good Thing®

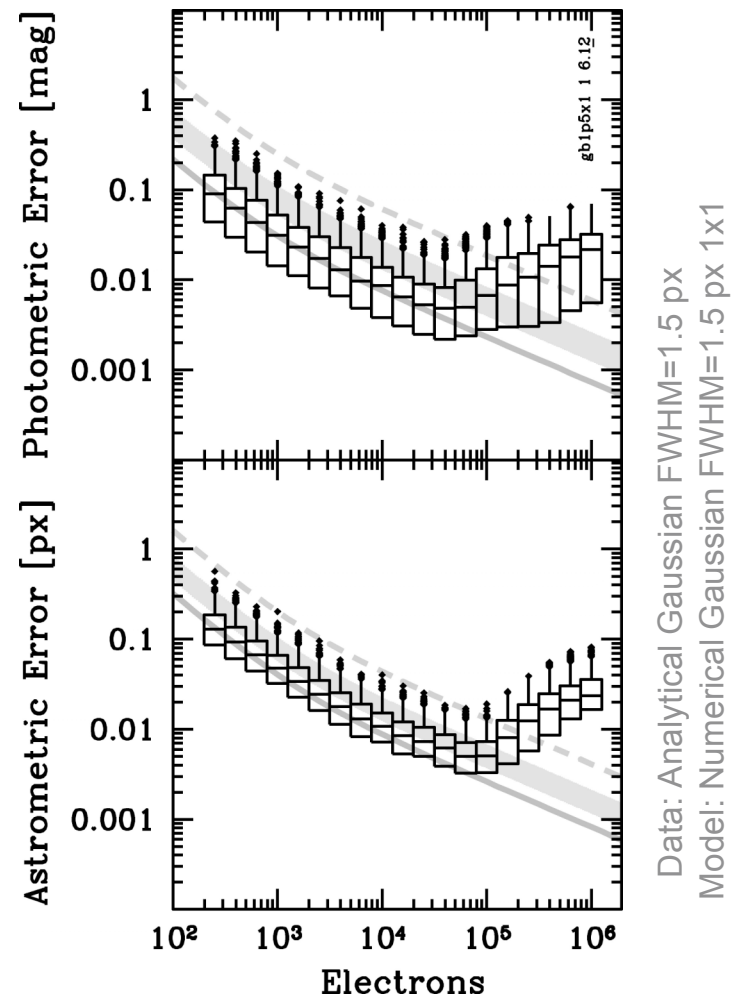


Computational noise < 1/15th photon noise

Inefficient Detectors

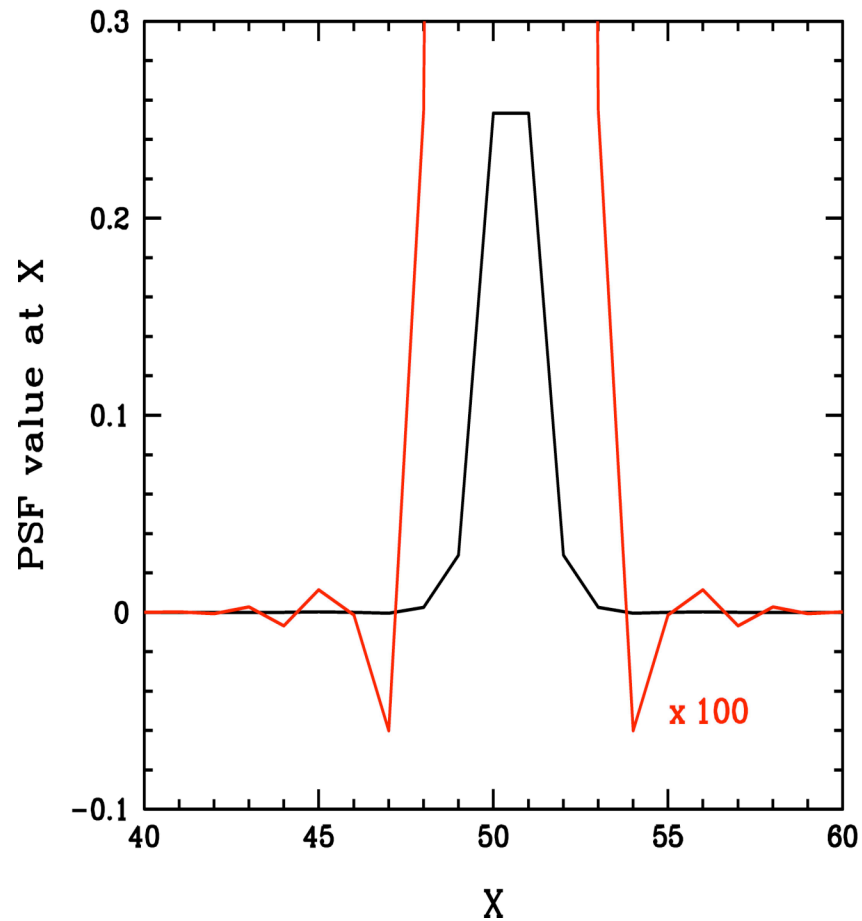


Problem: Undersampled PRFs



Source of the problem: *Negative flux values* caused by shifting an undersampled PSF !

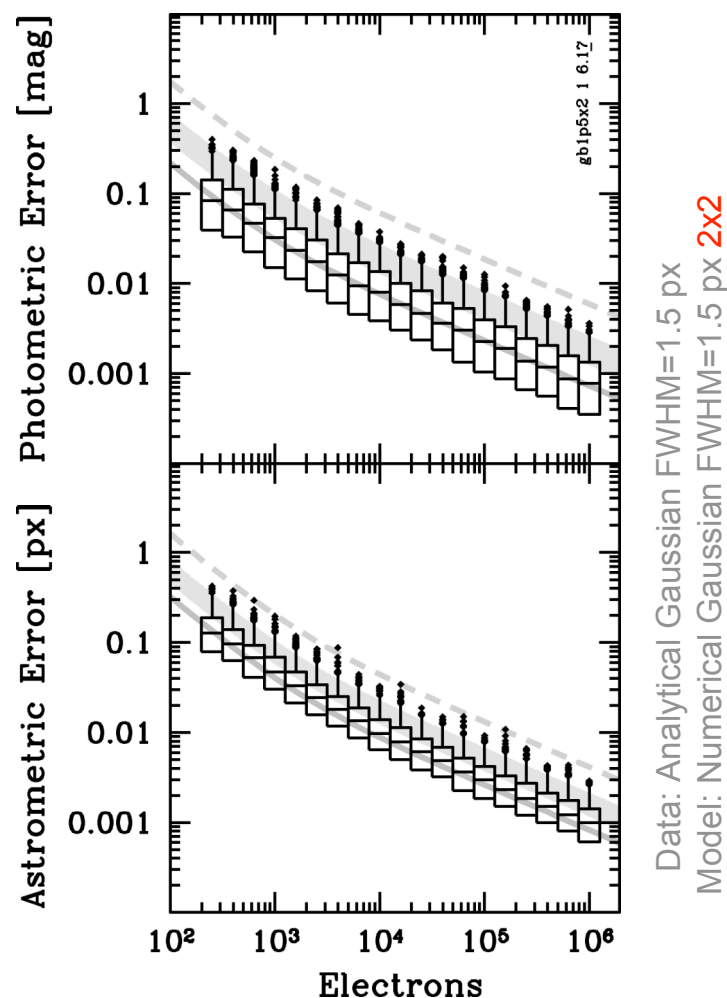
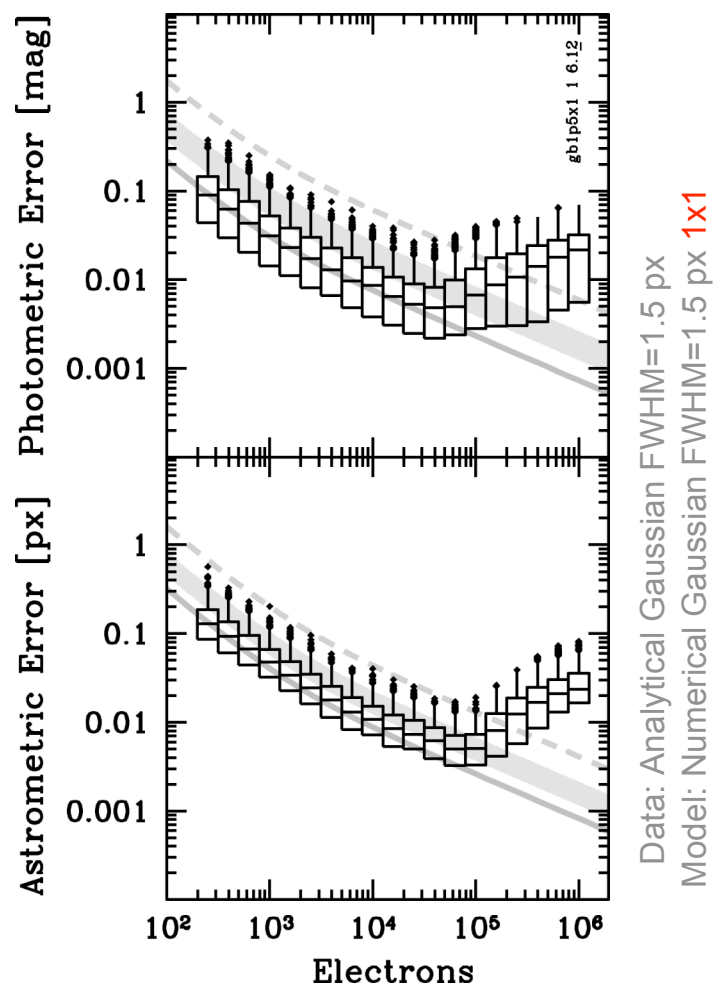
A pixel-centered Gaussian with FWHM=1.5 shifted 0.5 px in X



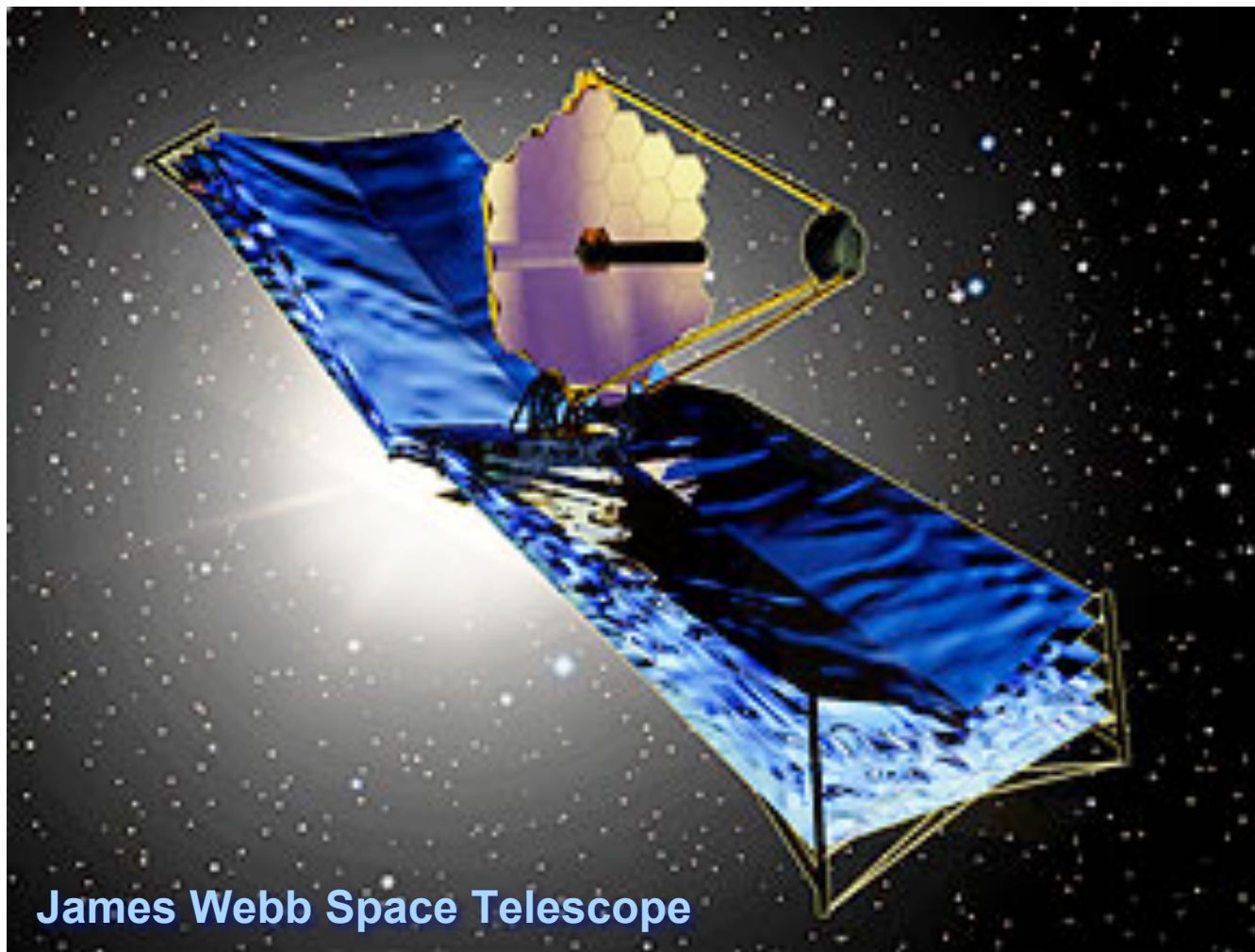
Question: But why did it work for faint stars?

Because one need lots of photons to properly sample the higher spatial frequencies of a PSF!

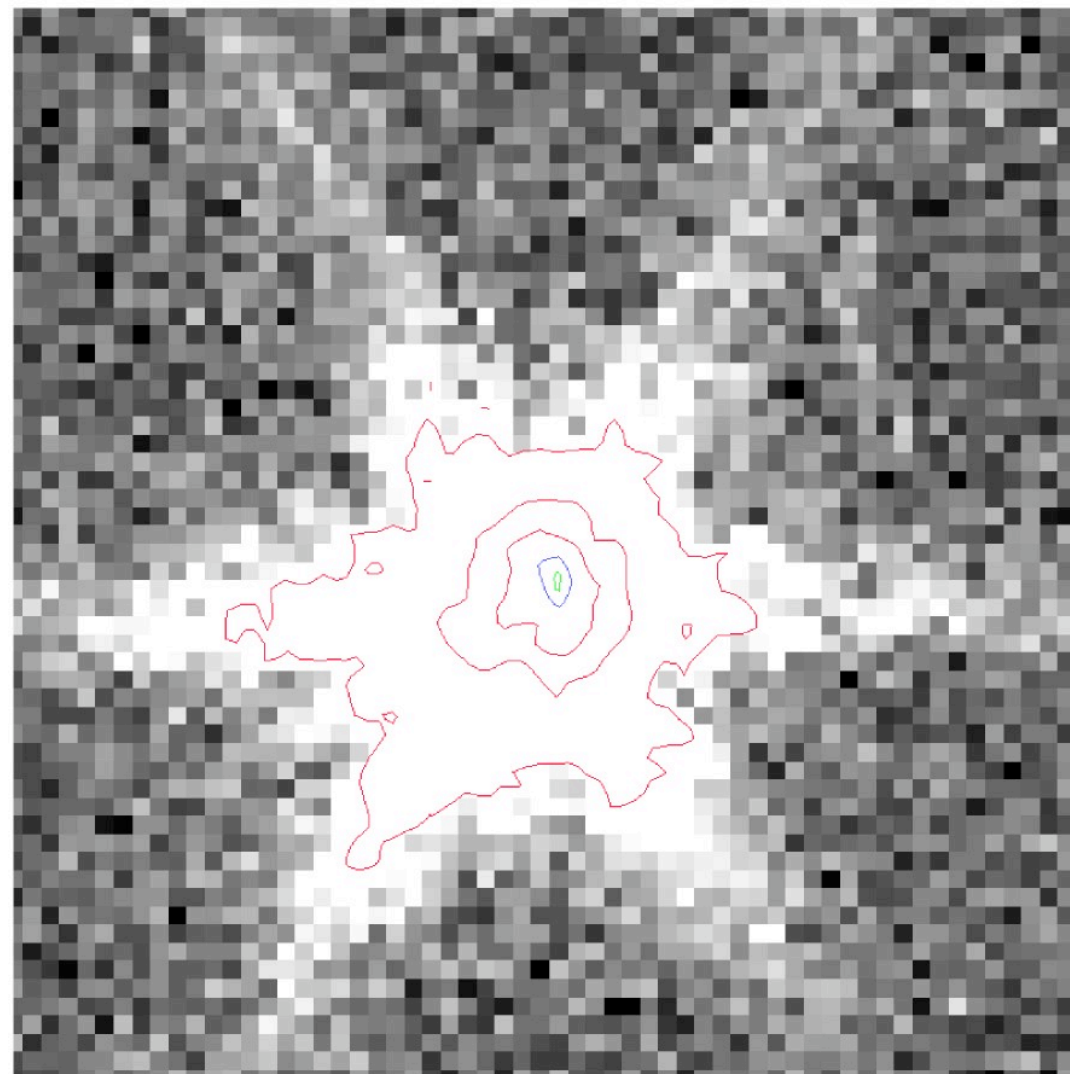
Solution: Use Supersampled PSFs !



Ugly (real) PSFs?

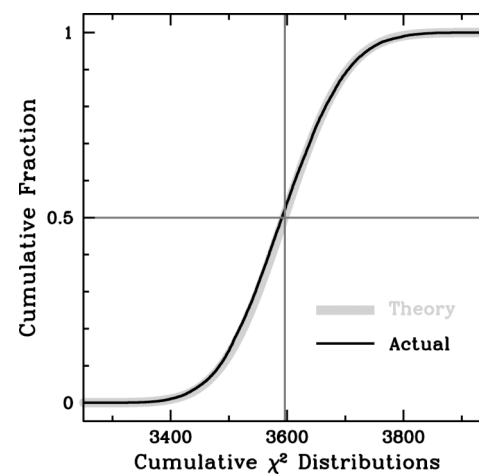
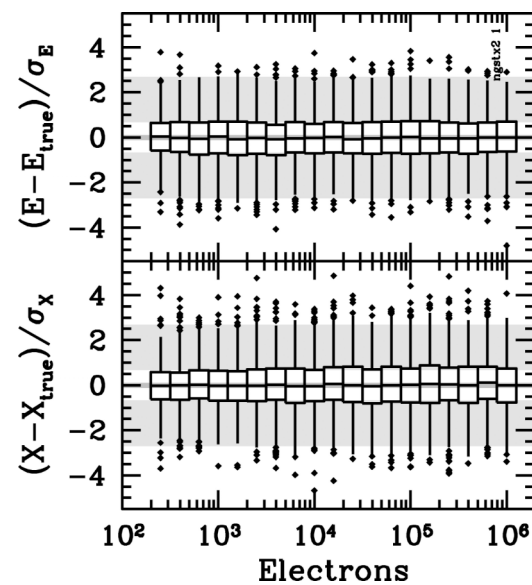
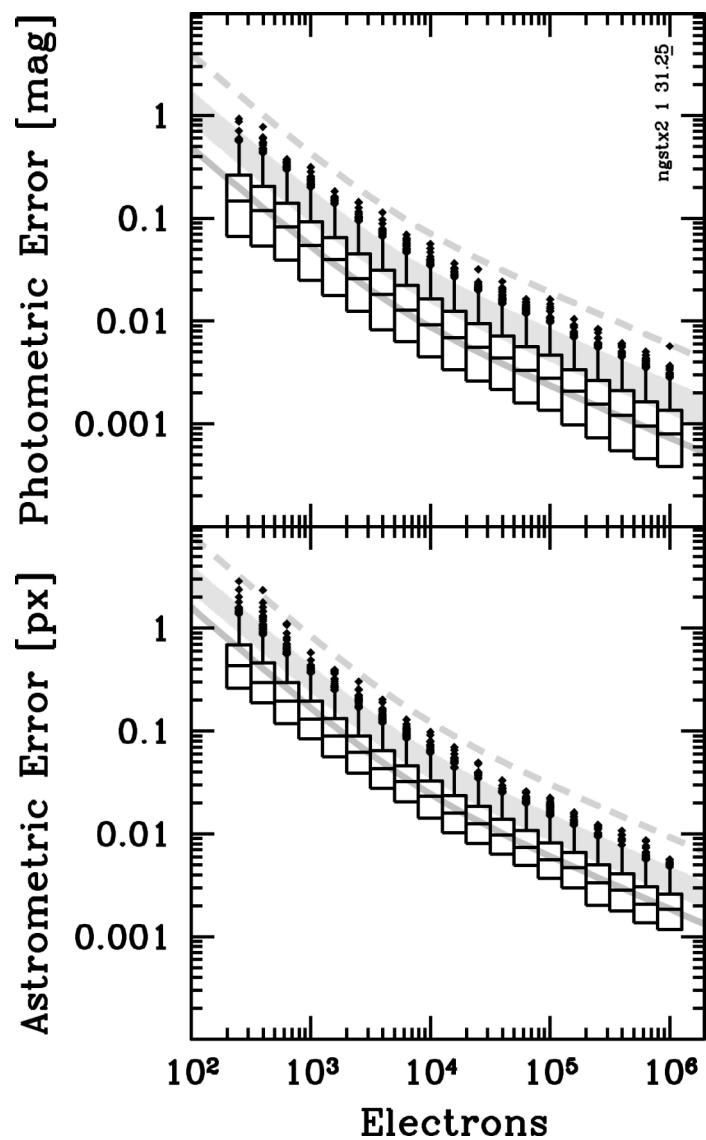


Next Generation Space Telescope 8-m TRW-concept



contours: 90%, 50%, 10%, 1%, 0.1% of peak

0.0128 arcsec/pixel $\beta=31.05$ PSF by John Krist



Calibration Errors



Challenges of PSF Extraction

- highly **variable PSF** within the field of view
- **too few stars**
- which are **not bright enough**
- from **undersampled observations**
- that are **poorly dithered**
- with
 - significant **Charge Transfer Efficiency** variations
 - **variable diffusion**
 - loss of photons due to **charge leakage**
- and may possibly be **nonlinear** at $<1\%$ level

Spitzer Space Telescope's Infrared Array Camera

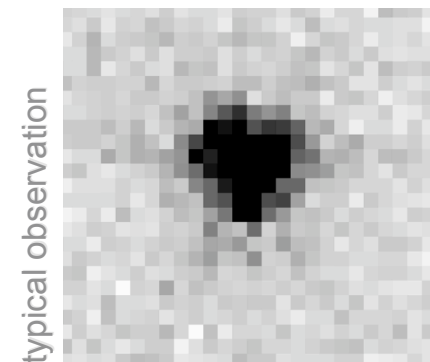
Infrared Array Camera

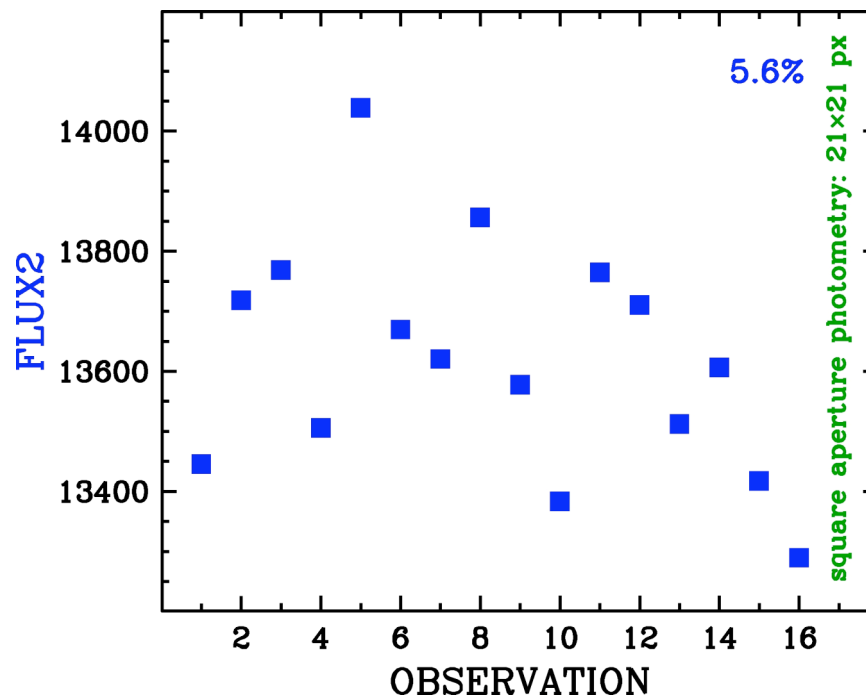


Spitzer Space Telescope

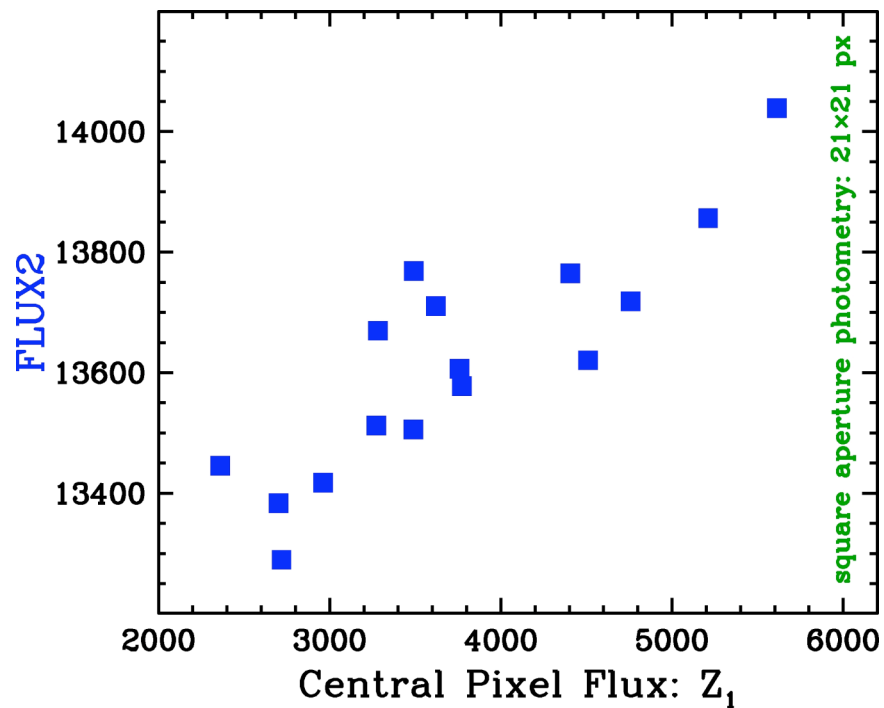


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3	17h06m10.8s	+73d40m10s	0.4	2003-10-08T12:22:01.538	ads/sa.spitzer#0006876928
4	17h06m10.6s	+73d40m09s	0.4	2003-10-08T12:35:06.524	ads/sa.spitzer#0006877184
5	17h06m11.3s	+73d40m12s	0.4	2003-10-08T12:48:11.510	ads/sa.spitzer#0006877440
6	17h06m10.9s	+73d40m12s	0.4	2003-10-08T13:01:16.496	ads/sa.spitzer#0006877696
7	17h06m10.5s	+73d40m11s	0.4	2003-10-08T13:14:21.489	ads/sa.spitzer#0006877952
8	17h06m10.2s	+73d40m11s	0.4	2003-10-08T13:27:26.471	ads/sa.spitzer#0006878208
9	17h06m11.0s	+73d40m14s	0.4	2003-10-08T13:40:31.472	ads/sa.spitzer#0006878464
10	17h06m10.7s	+73d40m13s	0.4	2003-10-08T13:53:36.446	ads/sa.spitzer#0006878720
11	17h06m10.5s	+73d40m13s	0.4	2003-10-08T14:06:41.436	ads/sa.spitzer#0006878976
12	17h06m10.0s	+73d40m12s	0.4	2003-10-08T14:19:46.422	ads/sa.spitzer#0006879232
13	17h06m11.0s	+73d40m15s	0.4	2003-10-08T14:32:51.423	ads/sa.spitzer#0006879488
14	17h06m10.5s	+73d40m15s	0.4	2003-10-08T15:06:39.788	ads/sa.spitzer#0006879744
15	17h06m10.3s	+73d40m14s	0.4	2003-10-08T15:19:44.785	ads/sa.spitzer#0006880000
16	17h06m10.0s	+73d40m13s	0.4	2003-10-08T15:32:49.763	ads/sa.spitzer#0006880256





A 5.6% peak-to-peak range in measured flux values within a 21x21 pixel square aperture.



Not all of the
variation is
random!

Stars centered in the middle of a pixel have more flux
than those that are centered on a pixel corner.

IRAC Data Handbook

Photometry and Pixel Phase - Ch1

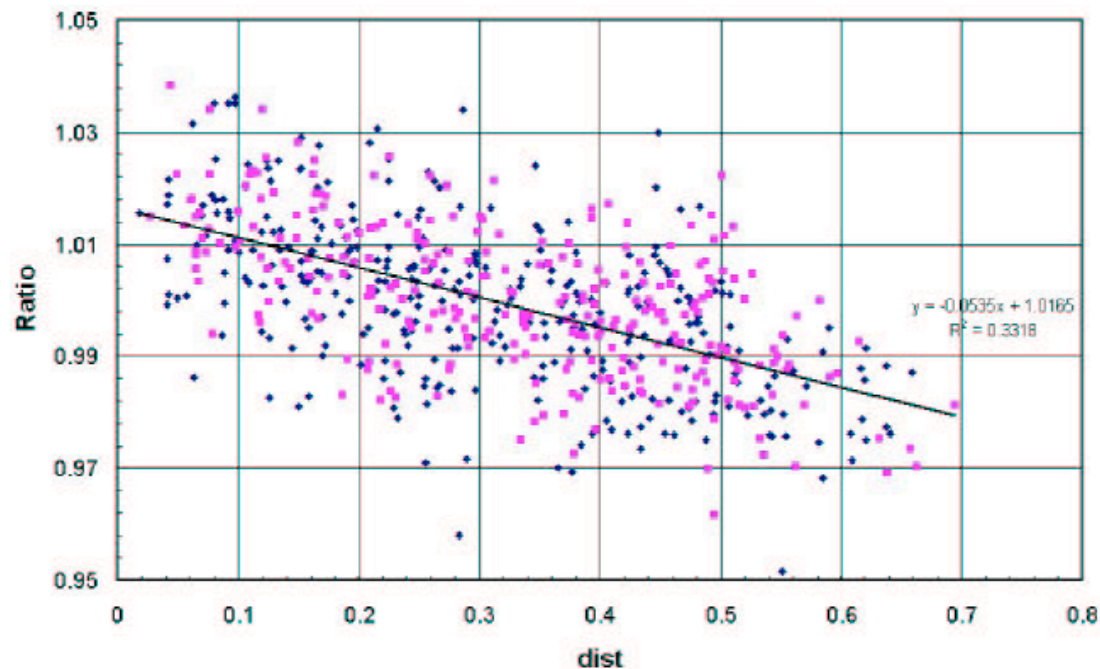
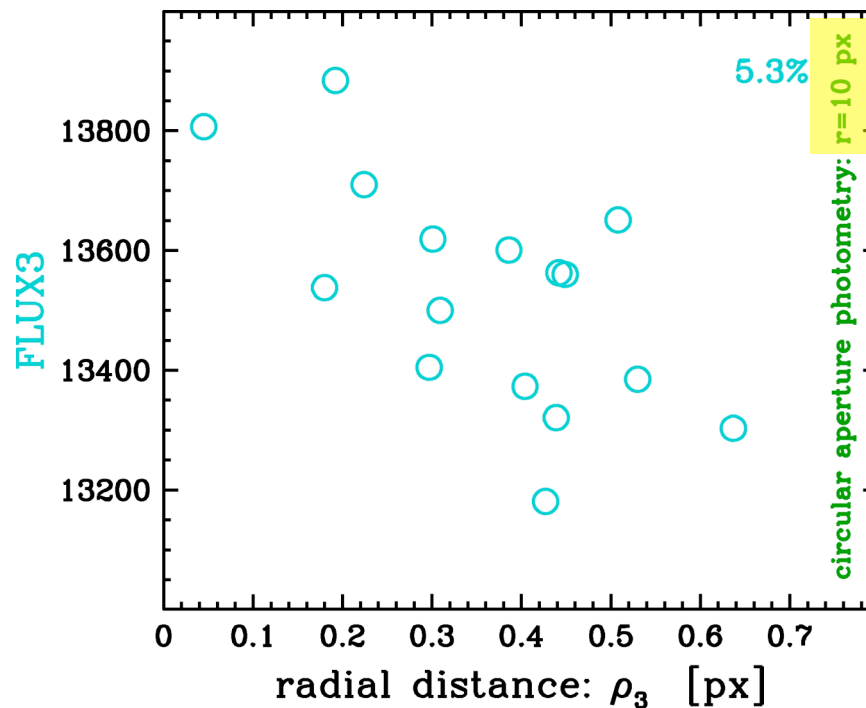
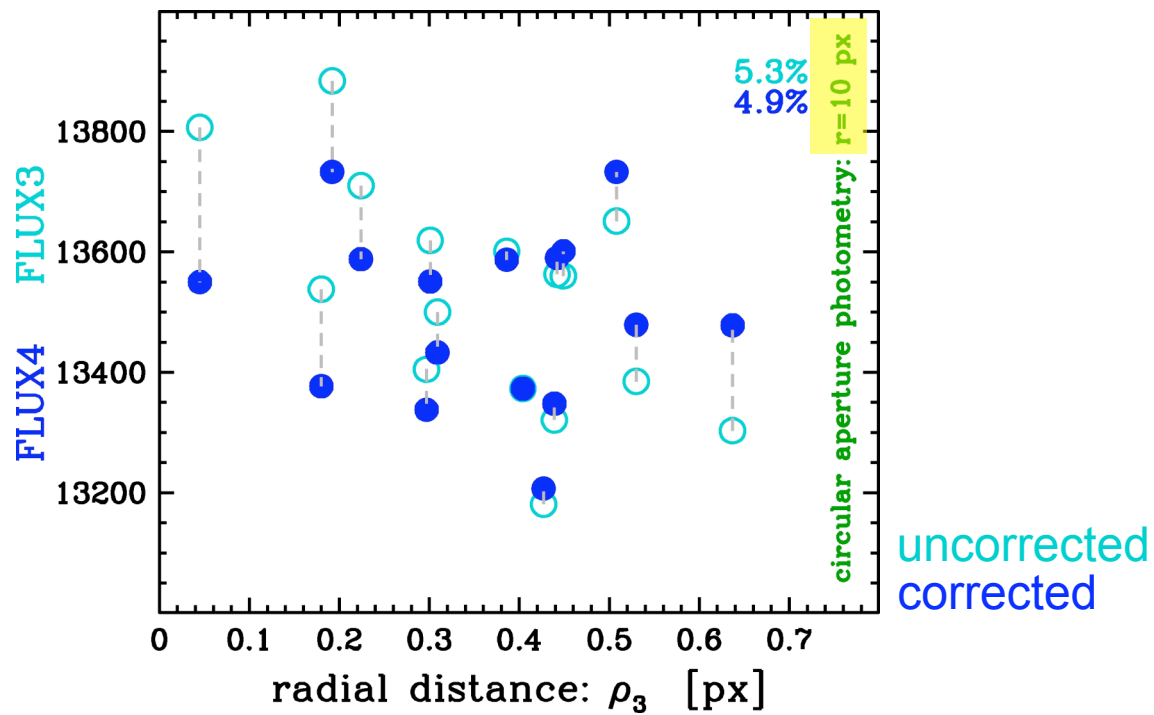


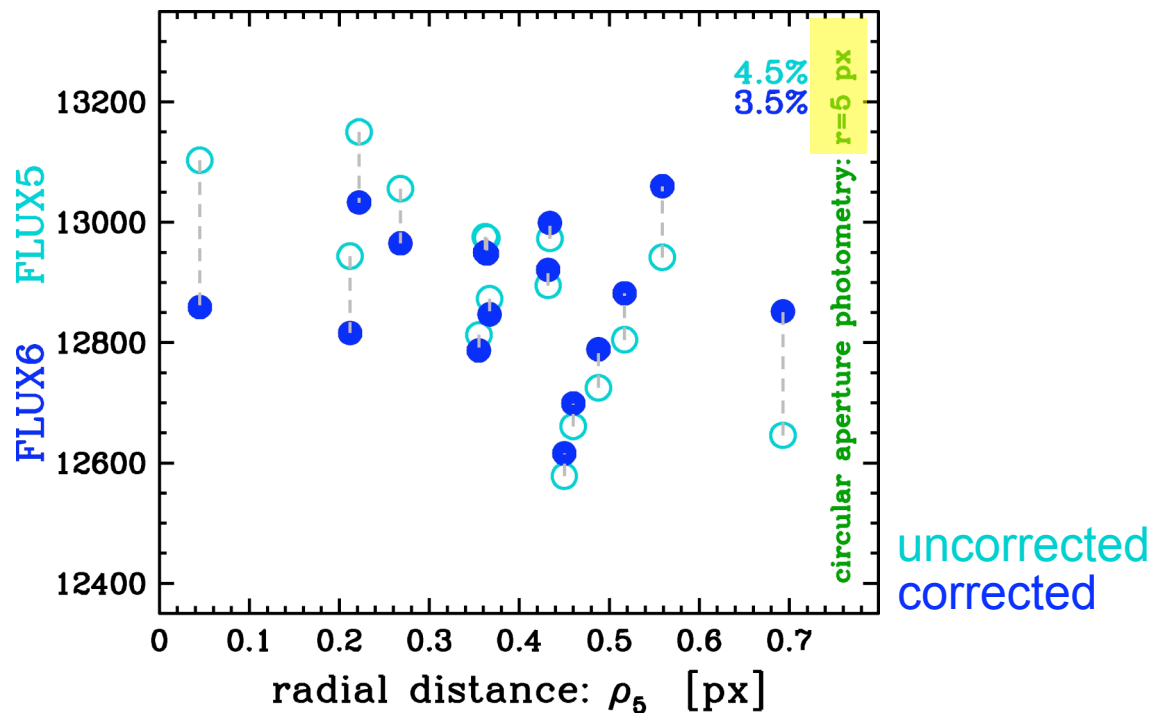
Figure 5.1: Dependence of point source photometry on the distance of the centroid of a point source from the nearest pixel center in channel 1. The ratio on the vertical axis is the measured flux density to the mean value for the star, and the quantity on the horizontal axis is the fractional distance of the centroid from the nearest pixel center.



A 5.3% peak-to-peak range in measured flux values within an aperture radius of 10 pixels.



A 4.9% peak-to-peak range using
the recommended radial correction.

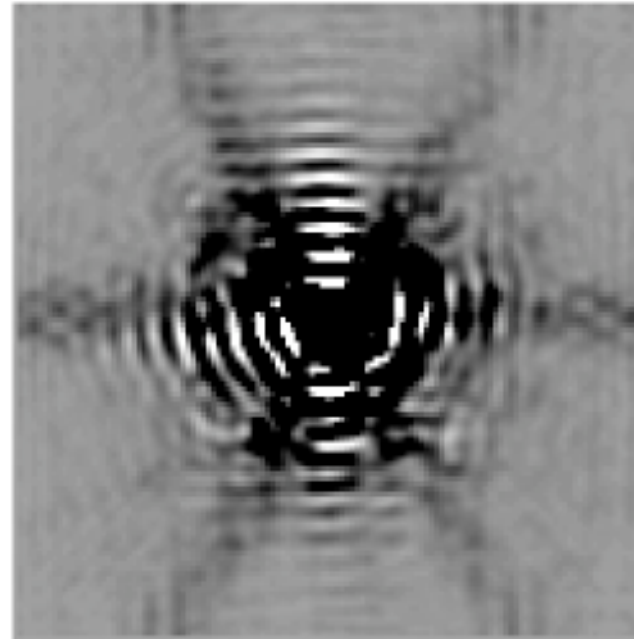


A 3.5% peak-to-peak range in measured flux values within an *aperture radius of 5 pixels* and using the recommended radial correction.

IRAC Ch1 PSF (5x5 theoretical)



linear stretch



logarithmic stretch

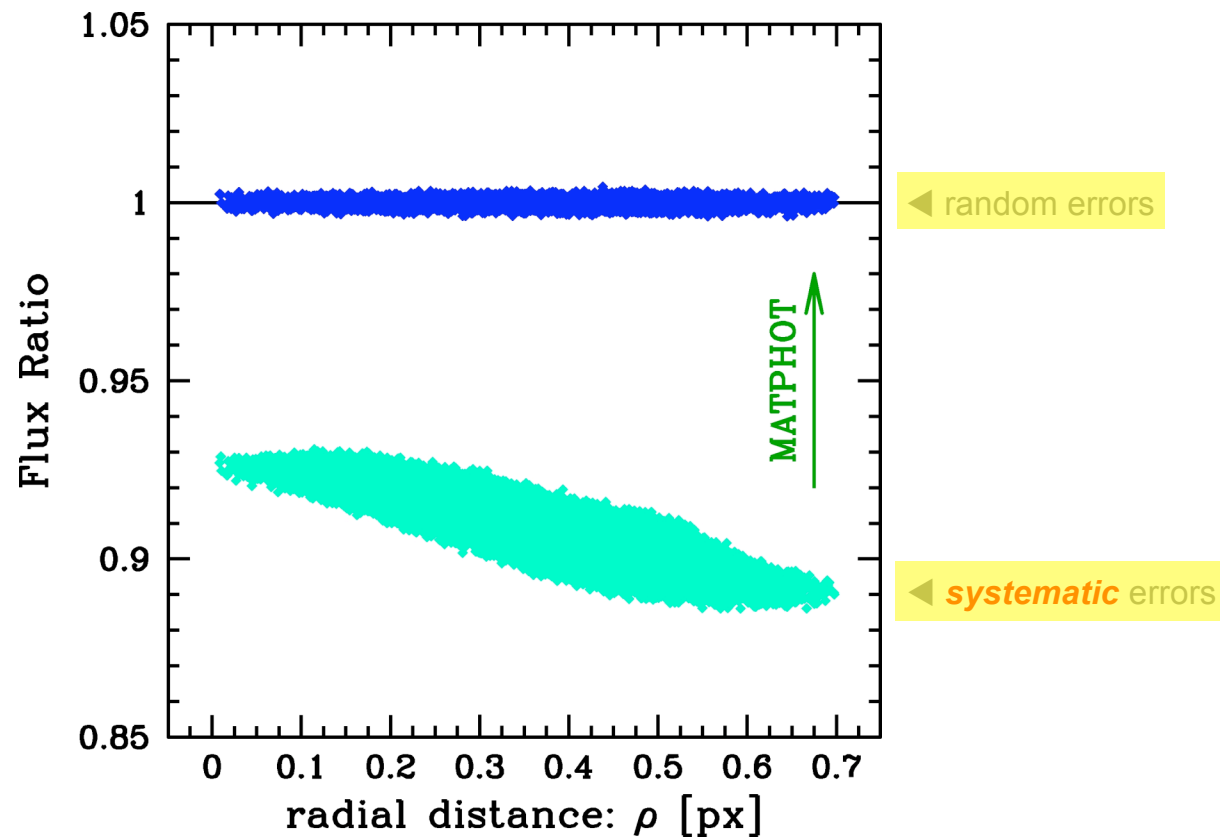
Source: Bill Hoffmann
(U. of Arizona, IRAC team member)

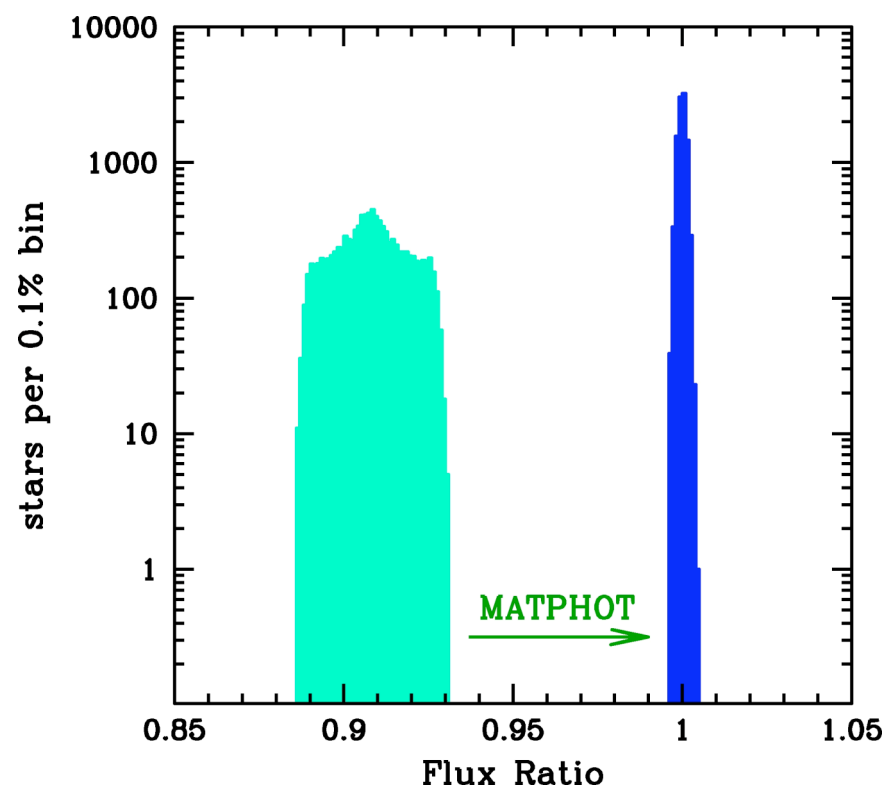
Relative Intrapixel QE Variation of IRAC Ch1

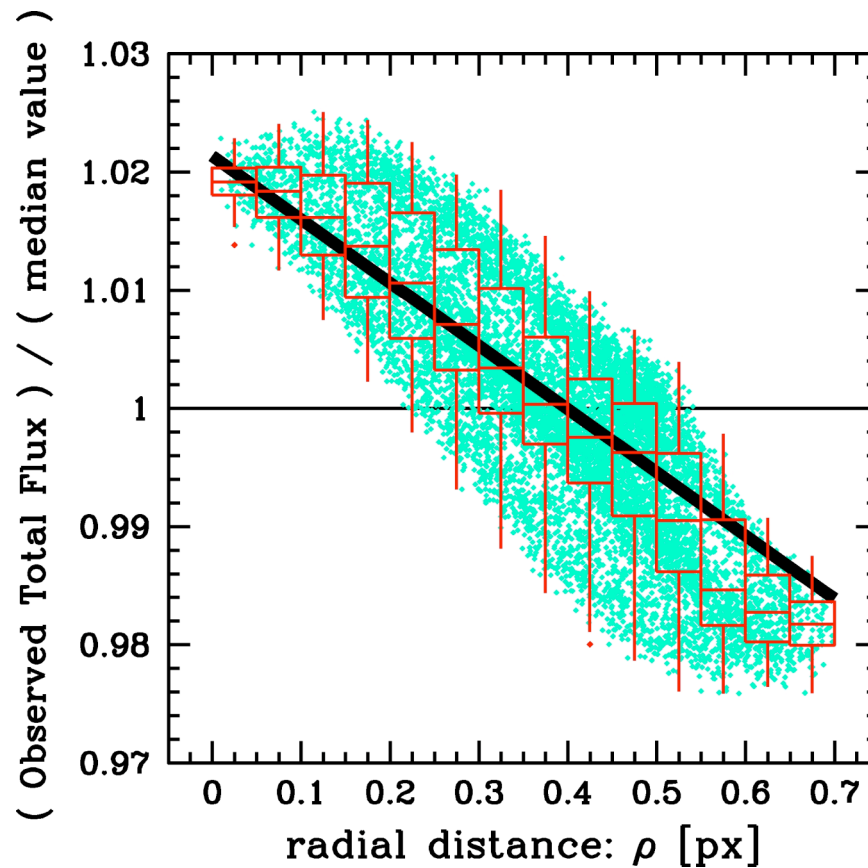
$$\text{intrapix} = \begin{pmatrix} 0.813 & 0.875 & 0.875 & 0.875 & 0.813 \\ 0.875 & 1.000 & 1.000 & 1.000 & 0.875 \\ 0.875 & 1.000 & 1.000 & 1.000 & 0.875 \\ 0.875 & 1.000 & 1.000 & 1.000 & 0.875 \\ 0.813 & 0.875 & 0.875 & 0.875 & 0.813 \end{pmatrix}$$

Source: Bill Hoffmann
(U. of Arizona, IRAC team member)

Lost flux can be recovered

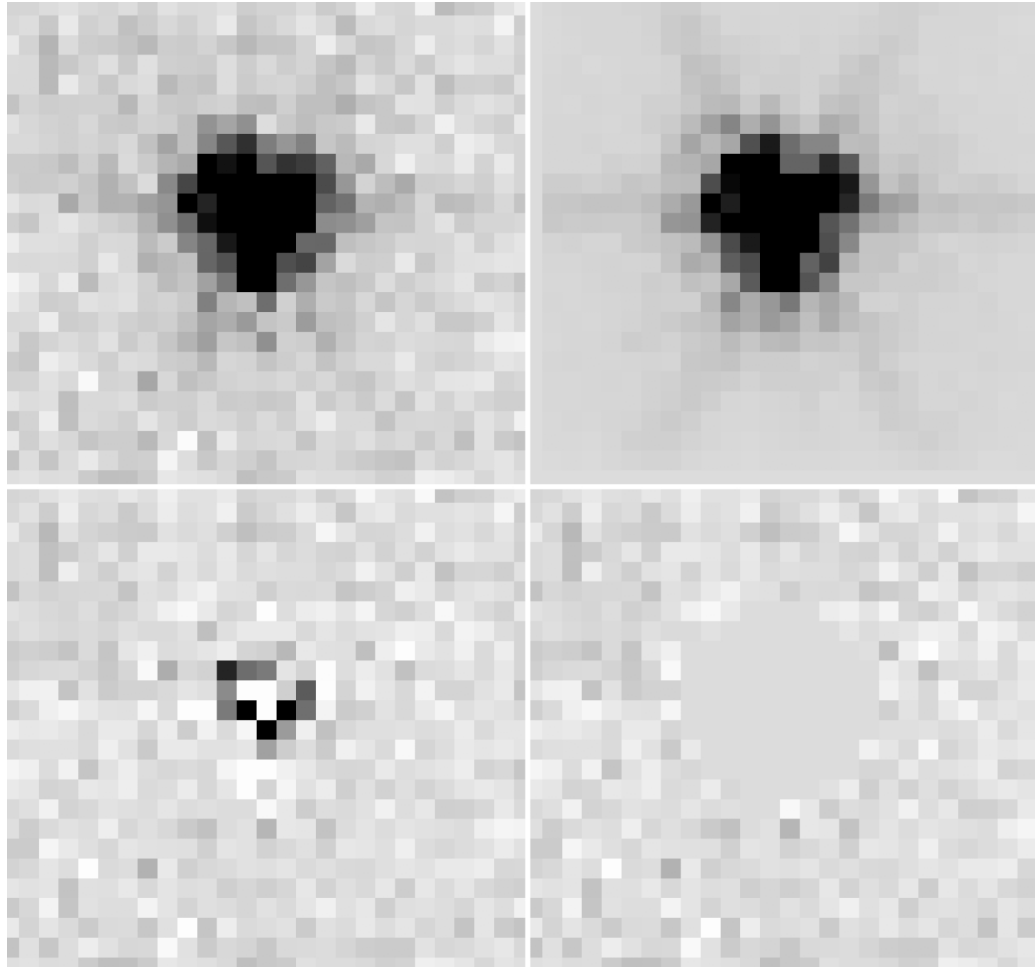


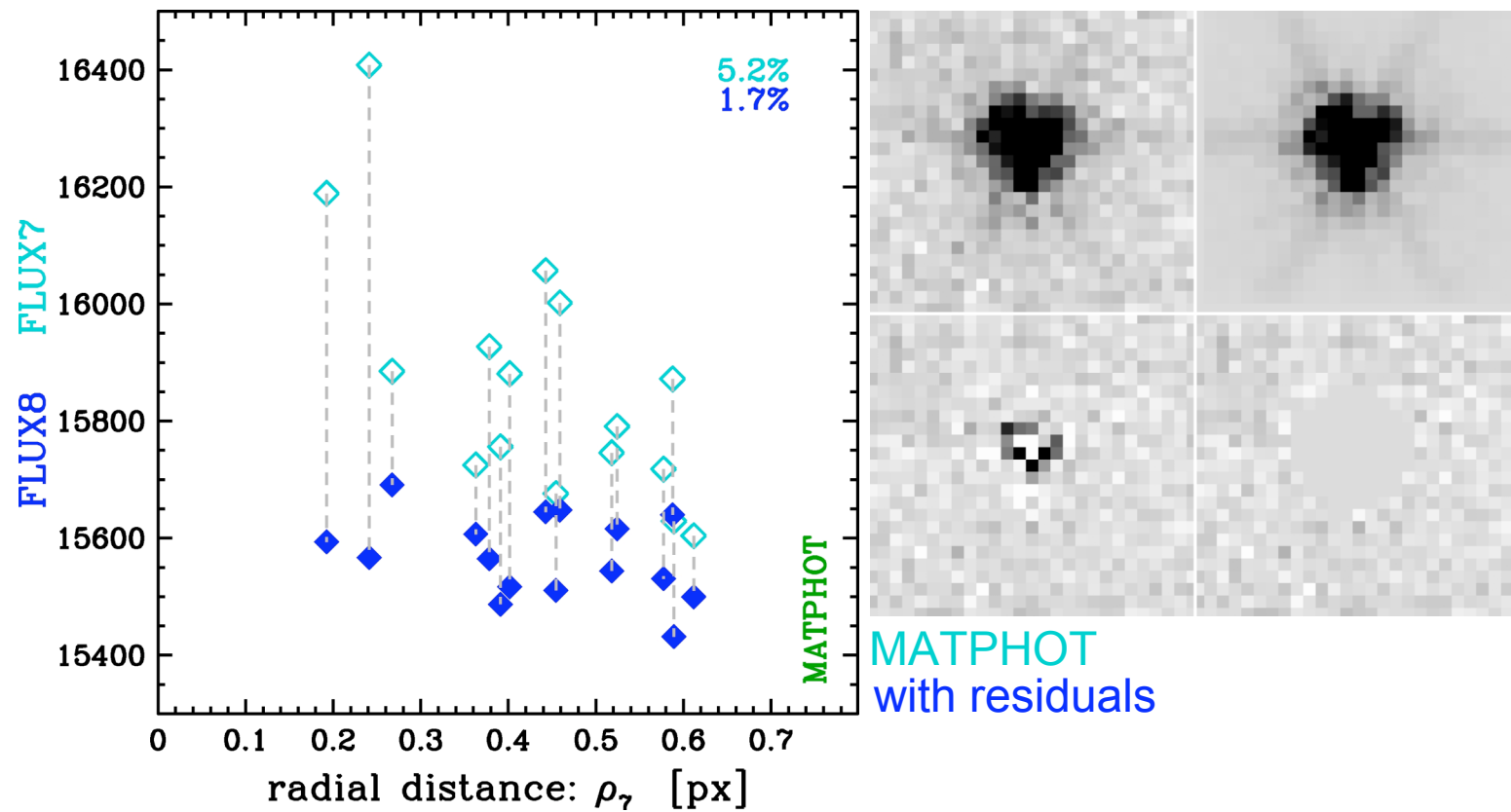




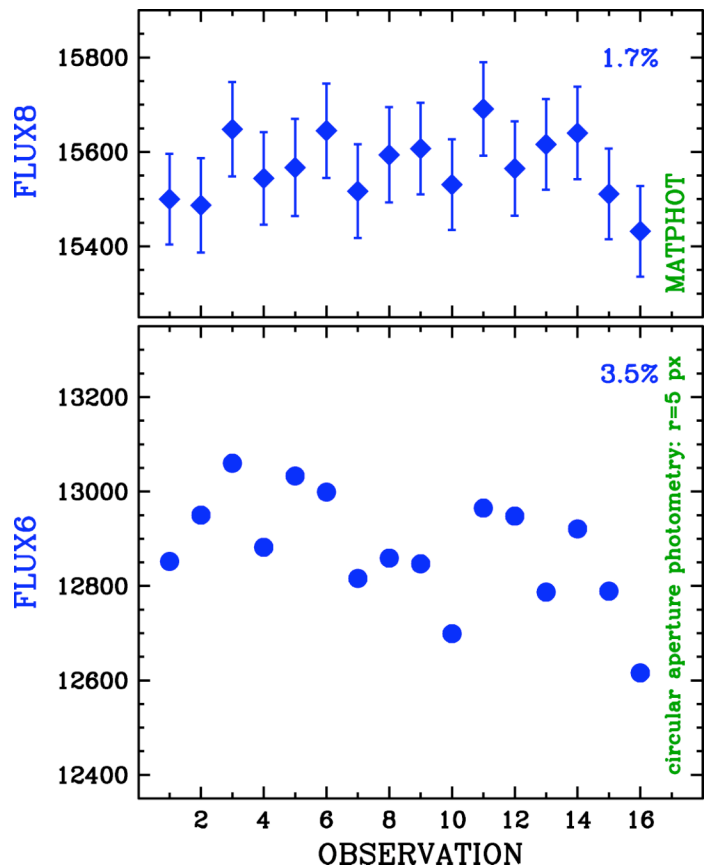
The recommended radial correction may leave a significant systematic error of a few percent.

MATPHOT Photometry



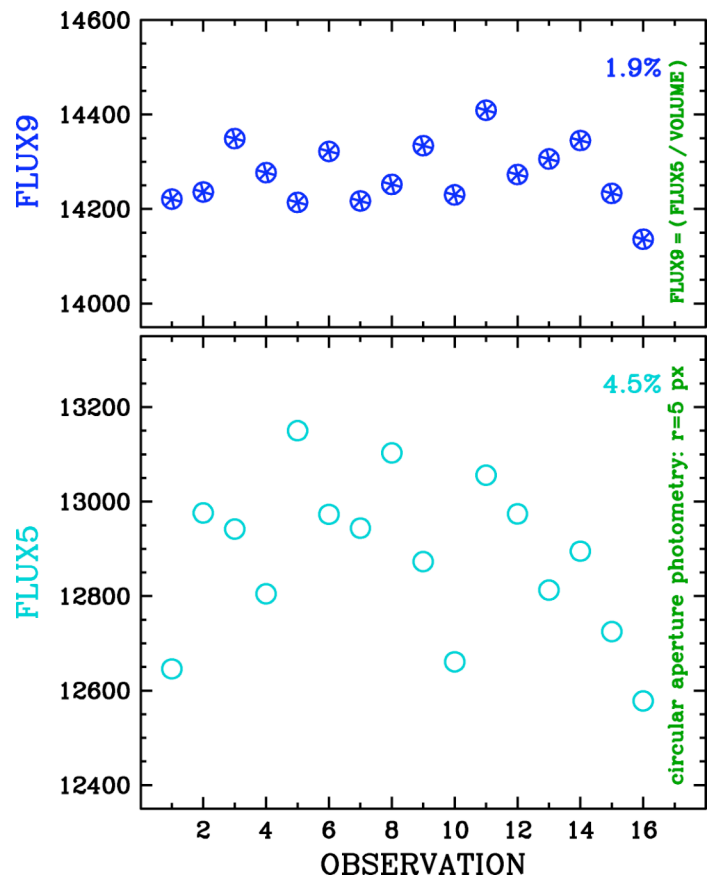


A 1.7% peak-to-peak range using MATPHOT flux measurements with residuals within 5 pixels.



An improvement of 100%
over the best results from
aperture photometry using
MATPHOT with residuals.

ID	X_7	Y_7	FLUX7	FLUX8	B_7	ρ_7	VOLUME	NOISEPX
1	132.529	125.391	15604 ± 96	15500	0.031	0.6118	0.889241	9.849
2	132.383	127.081	15756 ± 100	15487	0.062	0.3912	0.911503	6.357
3	132.277	128.366	16002 ± 100	15648	0.065	0.4589	0.901929	7.863
4	132.448	129.260	15746 ± 98	15544	0.088	0.5181	0.896910	8.521
5	131.197	125.860	16408 ± 103	15567	0.140	0.2412	0.925153	4.755
6	131.026	127.442	16057 ± 100	15645	0.096	0.4426	0.905820	7.435
7	130.971	128.599	15881 ± 99	15517	0.086	0.4017	0.910432	6.489
8	130.866	129.862	16189 ± 101	15594	0.092	0.1925	0.919395	5.478
9	129.737	126.250	15725 ± 97	15607	0.136	0.3629	0.898101	8.424
10	129.669	127.473	15718 ± 96	15531	0.146	0.5774	0.889749	9.571
11	129.795	128.172	15885 ± 99	15691	0.170	0.2673	0.906094	7.173
12	129.623	129.971	15927 ± 100	15565	0.174	0.3783	0.908991	7.253
13	128.649	125.610	15791 ± 96	15616	0.123	0.5244	0.895622	8.549
14	128.370	127.543	15872 ± 98	15640	0.078	0.5876	0.898904	7.922
15	128.627	128.259	15676 ± 96	15511	0.112	0.4544	0.894073	9.229
16	128.664	129.484	15629 ± 96	15432	0.102	0.5893	0.889795	9.532



corrected
uncorrected

An improvement of >100%
for *aperture photometry*
corrected with MATPHOT
computed PRF volumes.

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